

Recession Probabilities for the Eurozone at the Zero Lower Bound: Challenges to the Term Spread and Rise of Alternatives

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This paper examines the recession probabilities for the Eurozone along four different dimensions: First, we identify the best performing indicators for a recession within the next 12 months based on 43 underlying single variables and their different transformations in a benchmark model. We find that a modified version of the yield curve incorporating the shadow interest rate removes the downward rigidity of the front-leg and restores part of the informational content of the term spread at the zero lower bound. However, the best performing single indicator of the benchmark model is Real M1 followed by the Purchasing Managers Index (PMI), the investment grade corporate bond spread and the Terms of Trade. Second, the paper establishes three submodels to increase the lead-time and the stability of recession models: (i) Monetary transmission channels via principal component analysis; (ii) Bivariate regressions to identify paramount combinations; (iii) Unstable surges vis-à-vis the Hodrick-Prescott trend to detect animal spirits and hawkish mistakes. Third, the analysis is extended over various forecasting horizons (6m, 18m and 24m). Fourth, the results are analyzed from the perspective of risk-affine and risk-averse investors.

JEL-Classification: E3, E32, E37, E5, E58

**Keywords:** recession probability, term spread, zero lower bound, ECB, monetary transmission channel, animal spirits

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July 11, 2018

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#### 1. Introduction

The literature on predicting the likelihood of a recession provides broad consensus on the term spread as a reliable indicator for the binary state of the business cycle. Almost two decades after the formation of the European Monetary Union (EMU), data on the currency bloc challenges conventional wisdom through several dichotomies: A rather classical economic cycle characterized by the build-up of current account imbalances and overheating before the Global Financial Crisis (GFC) followed by a lackluster economy lingering in a liquidity trap and secular stagnation thereafter (Summers (2014)). Different monetary regimes ranging from orthodoxy and related hawkish mistakes were followed by unconventional monetary policy with a main refinancing rate pinned down at the zero lower bound (ZLB) and a neutral interest rate that potentially has declined substantially over time (Laubach and Williams (2015)). In addition, the brief history of the EMU is unique compared to other currency areas and the cohesiveness of the underlying economic region has been hostage to politicians, monetary policy makers and financial markets. The re-pricing of risks in periods of substantial convergence of EMU countries and increasing frictions after the GFC has affected financial market variables beyond their predictive value vis-à-vis the business cycle. These particularities of the Eurozone require that predictors of recession and expansion phases need to be sensitive enough to detect different types of economic cycles, insusceptible to varying central bank policies, and robust enough to withstand risks of re-denomination within the currency area.

The main purpose of this paper is to assess the likelihood of a recession in the Eurozone within the next 12 months along four dimensions: The first dimension focuses on the selection and the relevant transformation of input factors in univariate probit models representing the benchmark model of the analysis. On a second dimension, the paper introduces three submodels: First, we establish monetary transmission channels via principal component analysis to remove idiosyncratic influences and increase the indicators' robustness. Second, we combine the most predictive indicators from the benchmark model to identify paramount combinations. Third, we aim at modelling animal spirits and hawkish mistakes by constructing *unstable surge* indicators based on the deviation of the time series to its Hodrick-Prescott trend. The third dimension extends the analysis to three additional time horizons to analyze the lead-time of the input factors for the 6m, 18m, 24m forecasting periods. The fourth dimension models the utility for risk-affine and risk-averse market participants based on a loss-function to identify their preferred choices.

The paper proceeds as follows: In section 2, a literature review sheds light on empirical findings. Section 3 discusses theoretical considerations on why the term spread might be prone

to issue ambiguous signals in the current environment. Section 4 discusses the methodology and the dataset of the analysis. Section 5 reviews the benchmark model which is extended to the various submodels in section 6. Section 7 examines various time horizons and presents the results of an out-of-sample analysis. The paper concludes with a summary of key findings in section 8.

#### 2. Literature Review

Given the binary classification of the business cycle into recessions and expansion phases, statistical procedures to predict the state of economic activity are inherently non-linear with probit models representing the most popular regression specification.<sup>1</sup> As the literature remains inconclusive about a superior type of modification<sup>2</sup>, the main pivots for raising the predictive power of forecasting models have been the choice of input factors and their transformations. Hence, this section will classify existing studies on recession probabilities into their findings on the best performing explanatory variables, their regional perspective as well as their performance across various forecasting horizons and different transformations.

Regarding the explanatory variables, the slope of the yield curve has been associated with paramount predictive power for the future state of the business cycle. According to Estrella and Hardouvelis (1991), a positive term spread is a leading indicator for an expansion of consumption, investment and production. These results are underpinned by more recent research by Chauvet and Potter (2005), Rudebusch and Williams (2009) and Fornari and Lemke (2010). While the term spread has delivered strong predictive power across different pivots, Moneta (2005) examines ten variations of the yield curve (attachment points: 3m, 1yr, 2yr, 5yr, 10yr) and identifies the 3m-10yr yield curve as the most explanatory specification. Notwithstanding its widespread use, the term spread shows certain instability as the momentum of its inversion in the run-up to a recession can vary significantly with the type of economic downturn.

The inclusion of additional variables in multivariate probit models yields improved results. King et al. (2007) show that bivariate models of the investment grade (IG) or high-yield (HY) corporate bond spread in combination with the yield curve reduce the occurrence of false positive signals significantly. Based on theoretical considerations of Gilchrist and Zakrajšek

<sup>1</sup> A few studies also include logit regressions that produce similar results with slightly fatter tails of the probability distribution.

<sup>&</sup>lt;sup>2</sup> Modifications to probit models yield partial improvements of results by including numerous states of the business cycle via multiple breakpoints in Markov-switching models (Dueker (1997), Levanon (2011), Chauvet and Potter (2005)), and VAR-augmented specifications (Fornari and Lemke (2010)).

(2012), Ergungor (2016) decomposes corporate bond spreads into two parts that reflect the expected default of the issuer<sup>3</sup> and a general excess bond premium measured as the difference between the corporate spread and the expected default premium. While the excess bond premium turns out to be highly significant in predicting recessions, the additional yield implied by expected defaults does not display statistical significance. Confirming the predictive power of credit spreads, Fornari and Lemke (2010) also ascribe stock returns the ability to augment the forecasting performance of the term spread. In an analysis to predict economic turning points, Dovern and Ziegler (2008) find that corporate spreads, survey data, and composite indicators of real economic variables increase the forecasting quality. However, the authors cast doubts on stock returns that issue multiple false signals. In a study by Liu and Moench (2014), the term spread is paramount over various time horizons and its performance increases most notably in combination with the 1yr change of the S&P 500, Broker-Dealer margin accounts, the ISM (Institute of Supply Management) new order index and the NAPM (National Association of Purchasing Managers) index of consumer commodity prices. An additional source of useful indicators stems from the universe of corporate financial data such as the dividend yield (Bellego and Ferrara (2009)) and pre-tax corporate profits (Ergungor (2016)) whose interpretation is vulnerable to the informationally dilutive effect of the USD exchange rate.4

While aforementioned variables possess rather unambiguous correlation with the business cycle, the literature also identified a number of factors whose sign seems to be dependent on the respective region. As the majority of studies on recession probabilities focuses on the U.S., some researchers extend their perspective to a multi-country panel (Fornari and Lemke (2010) and Haltmaier (2008)), and only a few focus explicitly on the Eurozone or selected EU countries (Moneta (2005) and Bellego and Ferrara (2009)). According to Haltmaier (2008), exchange rates are broadly significant but the sign of their coefficients is ambiguous. While a real depreciation has mostly signaled a loss of confidence for developing countries (for example Mexico) coinciding with negative growth rates, in contrast a real appreciation has increased the recession probability for the UK and Taiwan whose open economies are more vulnerable to a loss of price competitiveness vis-à-vis the world. The same study shows that oil prices are significant for a selection of countries (U.S., the UK, Korea, and Taiwan) and insignificant for Canada and Mexico. A theoretical explanation lies in the cost-push effect for oil importing countries and the corresponding income-enhancing effect for oil-exporting

<sup>3</sup> The expected default rate is calculated based on historical sensitivities.

<sup>&</sup>lt;sup>4</sup> As the USD serves as a flight to quality currency, the J-curve effect could lead to an initial increase in the sales value despite a deterioration of the economic environment.

nations. Variables falling short of predicting recessions have been gold prices, silver prices, treasury rates (King et al. (2007)) and broader monetary aggregates (except for M1). In particular, the forecasting power of real M2 has been in retreat since the late 1980s (Levanon et al. (2011)). The authors even identify a relationship that is inverse to the assumed causality with an increase of real M2 during the recession of 2001 and the GFC, potentially caused by de-regulation and structural changes of the banking sector.

The forecasting horizons covered by literature range from now-casting to up to 24 months. While results deteriorate for longer horizons, the predictive power of variables is the strongest for forecasting periods up to 12 months. However, shortening the forecasting horizon does not necessarily lead to improved results as the key determinant is the chronological relationship of the variable with the business cycle. Explanatory variables enter into statistical models as levels, changes over certain time periods (mostly 6m or 12m change) and lagged values. Procedures to reduce the dimensions of datasets have become increasingly popular. Predominantly based on principal component analysis, the use of indices as input factors yields the benefit of at least partly removing idiosyncratic characteristics of the underlying dataset (Levanon et al. (2011) and Chen et al. (2011)).

#### 3. Conceptual Challenges to the Term Spread at the ZLB

From a conceptual perspective, the yield curve draws its explanatory value from two reinforcing sources: First, the inversion of the yield curve can be interpreted as a downward revision of market expectations on future real and nominal interest rates in the light of a decelerating economy. In line with the market expectations hypothesis, a decline of the long end of the interest rate curve reflects lower inflation expectations and dovish central bank policy that will be priced in at the first signs of a looming recession.

The second source of a flattening of the yield curve is monetary tightening in the boom phase of an economic cycle in order to contain inflationary pressures. As central banks have engaged in pre-emptive tightening to tackle inflation, short-term interest rates tend to rise faster than nominal interest rates in the mid-to-late cycle stage of an expansion, leading to a flattening of the curve.

At the ZLB, however, the shape of the yield curve as a predictor of the future state of the business cycle is compromised from two angles: The increasing use of forward guidance by central banks and a higher importance of the market segmentation theory in light of massive central bank asset purchases. Regarding the first aspect, central banks have indicated that future short-term interest rates would remain subdued for a pro-longed period of time. While this

forward guidance aimed at providing economic stimulus to escape the liquidity trap (Krugman (1998)) given dysfunctional fiscal and economic policy, markets incorporated the outlook on future policy rates into declining forward rates. While the conventional calibration of recession models using the term spread would be signaling elevated probabilities of an economic downturn, the cause for the flattening is actually a new monetary instrument in a low-growth environment.

The second differentiating characteristic at the ZLB is higher importance of the market segmentation hypothesis. While certain investors such as foundations, retail clients, hedge funds are less bound to certain maturity segments and asset classes, other investor groups like pension funds, insurance companies and foreign central banks are reliant on government bonds and certain maturity segments that serve their quality, liquidity and asset-liability matching needs. As these investors tend to stick to their preferred habitat, asymmetric bond purchases of certain maturities by a central bank create downward pressure of corresponding yield levels without a rotation out of these segments for higher risk-adjusted returns.<sup>5</sup> Given that the main refinancing rate is already at the ZLB and the deposit facility faces a natural lower bound, both asymmetric and symmetric bond purchases will have a stronger effect on the long end of the curve than on the front-end.

#### 4. Model Set-up and Data

#### 4.1 Model Set-up

Research on recession probabilities refers commonly to the standard probit model specification to capture the binary state of the dependent variable and the continuous nature of input factors:

$$P(Y_{within(t+k)}) = \theta (\alpha_0 + \alpha_1' X_t),$$
 [1]

where Y describes the binary state of the business cycle within the next t+k months. If one of the next 12 months is defined as a recession month, Y takes the value of 1, otherwise, it is 0. The probit model is based on the standard normal cumulative density function  $\theta$ . Thus, the model assumes non-linear behaviour of recession probabilities.  $\alpha_0$  denotes a constant term and  $\alpha_1$  represents the coefficients for the vector  $X_t$  which contains the input factors. In order to avoid over-specification and to be parsimonious, the following analysis only estimates

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<sup>&</sup>lt;sup>5</sup> In particular, the repo market is dependent on collateral and the scarcity in government bond markets puts downward pressure on yields without containing predictive quality regarding upcoming recessions.

univariate and bivariate models. As compared to recession models that forecast the probability of a downturn "in" 12 months, this analysis is more useful in detecting the actual length of a recession at the expense of the leading nature of the model. As a remedy for this statistical trade-off, the analysis explicitly considers the 6m, 18m and 24m forecasting horizons to vary the lead-time of the calibration.

The baseline model of this paper covers a univariate specification of the standard probit regression and considers three transformations of the input factors (the level, the 6m change and the 12m change).<sup>6</sup> In addition, the analysis establishes three submodels: The first submodel uses monetary transmission channels as input factors derived from a principal component analysis; the second submodel focuses on identifying superior bivariate model specifications; the third submodel aims at detecting unstable surges of economic developments to increase the lead-time of recession signals. The exact set-up of each submodel will be explained in greater detail at the onset of sections 6.1, 6.2 and 6.3.

#### 4.2 Data

The analysis covers monthly observations for the time period between December 2001 and December 2017 based on data from Bloomberg and Haver. The binary dependent variable relies on the quarterly classification of the CEPR Euro Area Business Cycle Dating Committee which indicates that 33 months out of a total of 193 are classified as recession months. The Committee defines a recession as "a significant decline in the level of economic activity, spread across the economy of the euro area, usually visible in two or more consecutive quarters of negative growth in GDP, employment and other measures of aggregate economic activity for the euro area as a whole." The concept to rely on a number of characteristics increases the resilience of the classification towards data revision and economic relevance.

As we focus on the probability of an economic downturn *within* a certain time period, the list of recession observations is modified to reflect the various forecasting horizons of 6m, 12m, 18m and 24m.<sup>8</sup> The underlying dataset for the 43 explanatory variables [see Table 1 for the complete list] can be broadly classified into eleven categories: interest rates, the term spread, money supply, competitiveness, sentiment, refinancing conditions, preference for risk, industrial data, inflation, commodities and the wealth channel.

<sup>&</sup>lt;sup>6</sup> If the concept of the indicator does not allow for stationarity, the level is not included in the analysis.

<sup>&</sup>lt;sup>7</sup> Given the limited history of the Euro area, we associate single months with the quarterly classification of the CEPR instead of aggregating the monthly input factor to quarterly periodicity.

<sup>&</sup>lt;sup>8</sup> For example, the definition of a recession "within six months" requires the binary variable to signal a recession at time t=0 if t+5 is a recession month, although the economy might expand in t+6.

#### < Insert Table 1 >

The first three groups represent traditional channels of central bank policy via direct control of short-term *interest rates* and the *monetary base* as well as indirect control of the *shape of the yield curve*. To account for the downward rigidity of interest rates at the ZLB, a modified version of the term spread considers a shadow short-term rate (Wu and Xia (2016)) that allows for reaching deeply negative territories. The remaining eight segments of variables extend the narrow view of conventional central bank measures to a more diverse group of monetary transmission mechanisms and various other influences.

Varying in their leading, coinciding or lagging nature vis-à-vis the business cycle, the time series are selected based on the existing literature and theoretical considerations about their predictive quality. In addition to the level of the respective variables, the study includes their 6m and 12m changes<sup>9</sup> to allow for different levels of sensitivity and to ensure reasonably mean-reverting characteristics in light of potential structural breaks.

#### 4.3 Measurement of Results

Following the recent shift in literature away from model evaluation based on fit indicators<sup>10</sup> to its actual classification ability, this paper assesses the quality of results from three different angles which are represented by a respective criterion: A signaling approach (Kaminsky and Reinhart (1999)), the area under the receiver operating curve (AUROC) and a loss function to model risk-averse and risk-seeking investors (Alessi and Detken (2009)).

The first criterion assumes that an indicator signals a recession when it surpasses the 50% threshold. Based on four categories, the signal is classified according to following matrix (Kaminsky and Reinhart (1999)):

	Recession within 12m	No Recession within 12m
Signal: Recession	A	В
Signal: No Recession	С	D

<sup>&</sup>lt;sup>9</sup> We either use the absolute difference between levels ("diff") or the %-change between levels.

<sup>&</sup>lt;sup>10</sup> To assess the quality of results, models have predominantly relied on fit indicators such as the Bayesian Information Criterion (King et al. (2007)), different versions of R<sup>2</sup> (Chen et al. (2011) and Moneta (2005)) and other loss functions like quadratic/logarithmic probability scores (Chauvet and Potter (2005) and Levanon et al. (2011)).

We will discuss indicators along a variety of performance ratios, including the number of recession months detected, false alarms and the overall percentage of correct predictions.<sup>11</sup>

As a remedy for the intuitive but somewhat arbitrary choice of the 50% threshold, this paper considers the area under the receiver operating curve (AUROC) as an alternative criterion. First introduced by Birdsall and Peterson (1953), this method was developed to express the diagnostic ability of a binary classifier system and has been recently used to describe the classification ability of business cycle indicators (Berge and Jordà (2011) and Liu and Moench (2014)). The centerpiece of this method is the receiver operating curve (ROC) which describes the ratio of true positive signals to false positive signals (i.e., false alarms) at various threshold values.

1. This method defines evenly spaced cut-off points, denoted as  $\Omega$ , which determine whether to interpret the indicated likelihood p as a recession or an expansionary signal. The narrower the cut-off points are, the smoother is the ROC curve. The example below assumes ten thresholds with a distance of 0.1.

$$\Omega = [0, 0.1, 0.2, \dots .0.8, 0.9, 1]$$

2. For each cut-off point, the process determines the ratio of true positive signals (TPS) to false alarms (FA).

$$TPS = \frac{1}{NR} \sum_{t=1}^{T} x$$
; where  $x = \begin{cases} 1, & p \ge \Omega \\ 0, & p < \Omega \end{cases}$ ; followed by a recession

where NR describes the number of recession months

$$FA = \frac{1}{NE} \sum_{t=1}^{T} z$$
; where  $z = \begin{cases} 1, & p \geq \Omega \\ 0, & p < \Omega \end{cases}$ ; followed by an expansion

where NE describes the number of expansion months

- 3. For each cut-off point, the procedure creates a coordinate in TPS-FA-space and connects these points to create the ROC. Coordinates above the 45% diagonal represent signals that are better than a random guess.
- 4. The AUROC is the integral of the area below the ROC.

The third criterion tries to shed light on the usefulness of indicators for different types of market participants following their relative preference of type 1 errors (missed recessions) versus type 2 errors (false alarm) along the lines of Alessi and Detken (2009). The one investor group that is rather sensitive to missing a recession will be labelled as *risk-averse*. The other

<sup>&</sup>lt;sup>11</sup> The definitions of the various performance ratios are given in the heading of the respective performance tables in the Appendix.

group of market participants, described as risk-affine, is more concerned with false alarms which would detract from performance in the expansion phase. The loss function L of both investor groups is defined as follows:

$$L = \rho \frac{c}{A+C} + (1-\rho) \frac{B}{B+D} .$$
 [2]

For the risk-averse investor group,  $\rho$  is assigned a value of 0.75 (which increases the loss from undetected recessions). For risk-affine investors, we assume  $\rho$  to be 0.25 (which causes higher costs associated with false alarms).

#### 5. Results of the Benchmark Model

This section highlights the challenges that the regime change of monetary policy at the ZLB is posing to the empirical supremacy of the term spread, suggests a remedy and then proceeds with a review of the best-performing indicators. Subsequently, the results are presented in a way that selected graphs displaying the probability level of indicating a recession for selected indicators (our first criterion) are included as figures (with panels for the selected indicators) in the main text body. The more detailed statistics on all indicators also including our second and third criteria are presented in corresponding tables in the Appendix [e.g., see Figure 1 and the corresponding Table A1 in the Appendix for our benchmark model].

While the recession of 2008-2009 still serves as a prime example for the ability of the term structure to forecast an economic downturn with sufficient lead-time<sup>12</sup>, the indicator fails to capture the second recession beginning in 2011 [see Figure 1a]. To the contrary, the signal inches north in the aftermath of the European Sovereign Debt Crisis and remains elevated for the time period between 2015-2017 in the wake of a flattening of the interest rate curve induced by the Expanded Asset Purchase Programme by the ECB starting in 2015.

In an attempt to restore the informational content of the yield curve, we suggest a modified version of the term spread [see Figure 1a] that incorporates the shadow rate by Xia and Wu (2016) and removes the downward rigidity of the front-leg.<sup>13</sup> As quantitative easing at

<sup>&</sup>lt;sup>12</sup> The term spread signals a recession probability of above 50% as early as November 2006

<sup>&</sup>lt;sup>13</sup> To have a better comparison to the traditional version of the term spread, we use the EURIBOR instead of the EONIA as an input factor to the modified version.

the ZLB can be perceived as a substitute for further rate cuts, this revised indicator is composed of a variable front-leg that switches from the 3m Euribor to the shadow rate as soon as the ECB' policy rate hits the ZLB.<sup>14</sup> Indeed, the ECB's unconventional monetary policy leads to a "shadow" steepening of the yield curve as the front-leg turns deep into negative territory. Hence, the average recession probability for the period between January 2015 and December 2017 recedes to 1% as compared to 40% as indicated by the traditional term spread.

The best performing indicator of the analysis is high powered money (12m change of Real M1), which reaches an almost perfect AUROC of 0.95, issuing 88% correct signals and detecting 73% of all recession months [see Table A1 in the Appendix]. In the run-up to both recessions, the indicator crosses the 50% threshold already 12 months before the start of each economic downturn [see Figure 1b]. Interestingly, Real M1 signals heightened recession risks around the near-crisis period at the beginning of last decade where economic growth had been lingering in negative territory for three months. Notwithstanding the fact that this time period did not technically qualify as a recession, these false positive signals underline the sensitivity of Real M1 to the economic cycle. Other aggregates of high powered money, M2 and M3, prove not to be useful as they are missing the leading character and are more exposed to other structural forces.

Slightly trailing the classification ability of Real M1, the second best performing cohort of variables is composed of indicators from sentiment indices, refinancing conditions, and measures of competitiveness. As the best indicator of the sentiment variables, the 12m change of the Purchasing Managers Index (PMI) [see Figure 1c and Table A1 in the Appendix] detects 60% of all crises with an AUROC of 0.88 and outperforms variables reflecting consumer or broader industrial surveys. Across all sentiment indices, the transformation to the 6m or 12m change has a significantly higher predictive value than the level due to a higher sensitivity to turning points. Regarding the recession of 2008-2009, the 12m change yields a lead time of 5m (3m for the second recession), while the level of the PMI first crosses a recession probability of 50% when the economy is already four months in recession (signal is coinciding with the start of the second recession).

Within the group of refinancing conditions, both spreads of investment grade (IG) corporate bonds and Eurozone peripheral sovereign bonds provide strong results [see Figure 1d and Table A1 in the Appendix]. Findings for the peripheral bond spread (12m change) indicate an AUROC of 0.85 and 49% of crisis detected. While the indicator is only somewhat flirting

<sup>&</sup>lt;sup>14</sup> The ZLB is defined as the situation where the lending facility is at 0% and the main refinancing rate hits 0.25% effectively limiting the space for further cuts (13th November 2013).

with the 50% threshold during the first recession, the variable issues strong recession signals as early as 16 months before the start of the second downturn, highlighting the causal relationship between tightening of financial conditions and economic growth in the Eurozone. This underlines the beneficial, dual characteristics of an indicator to reflect both a deteriorating endogenous business cycle and being itself a source for an exogenous shock. In light of the structural break in the wake of the Euro area crisis<sup>15</sup>, it turns out to be preferable to use the differenced time series as model input instead of the level. Particularly from the perspective of a risk-affine investor, this variable seems to be of use as it is less prone to non-systemic fluctuations with only 3% false positive recession signals [see Table A1 in the Appendix].

The IG corporate bond spread (level) yields an AUROC of 0.85 and detects 56% of recession months while issuing 7% false positive signals [see Table A1 in the Appendix]. The underperformance of high yield (HY) corporate spreads lends support to the critique of Gilchrist and Zakrajšek (2012) who suggest that the excess bond premium, and not the default risk premium, is the informative component about the future state of the economy. Since HY spreads are influenced to a stronger degree by the idiosyncratic default risk component, the actual sentiment among investors is reflected to a lesser degree.

In the *competitiveness* category, Terms of Trade detects 60% of crisis months with an AUROC of 0.85, moderately outperforming the FX EUR/USD rate (49% detected crisis and AUROC of 0.84) [see Figure 1e and Table A1 in the Appendix]. The strength of this indicator stems from the ability to include various characteristics such as the trade-weighted FX rate, import and export elasticities and the relative price of exports and imports. Periods of deterioration in the Terms of Trade can be a consequence of several events: a) shifts of global liquidity out of the EUR into USD-denominated save haven assets b) higher commodity prices increasing production costs c) lower export prices depressing internal consumption and investments. All these potential drivers are associated with a decrease in real economic growth and the probit model, hence, considers a significant deterioration of the indicator as a harbinger of a recession. The indicator appeals in particular to risk-averse investors who accept a relatively high number of false alarms of 13% in exchange for the third highest number of recession months detected.

Results for the other categories are overall mixed with some noteworthy exceptions. Out of the indicators reflecting *Preference for Risk*, the level of implied interest rates volatility yields an AUROC of 0.79 and detects 51% of all crises [see Figure 1f and Table A1 in the

<sup>.</sup> 

<sup>&</sup>lt;sup>15</sup> The first half of the last decade was characterized by a convergence process of sovereign bond yields and the assumption by capital markets that government debt of the Eurozone countries carried an implicit guarantee.

Appendix]. While the indicator is only coinciding with the first recession, the variable remains elevated in the preceding 14 months<sup>16</sup> before the second economic downturn capturing the uncertainty in financial markets during the peripheral crisis of 2011-2013.

Two other indicators that are sensitive to investor's varying preference for risk are the Libor-OIS spread (AUROC of 0.82 and 53% detected crisis) [Figure 1f and Table 1A in Appendix] and the EUR/USD FX Basis<sup>17</sup> (AUROC: 0.78 and 38% detected crisis; see Table A1 in the Appendix)). As a non-EUR specific indicator, the Libor-OIS spread accurately signals the GFC with a lead of eight months. On the flipside of its global nature, the variable is only coinciding with the second economic downturn which was idiosyncratic to the Eurozone. In the end of 2016, however, the indicator reveals its vulnerability to dynamics surrounding the infrastructure of the interbanking market: The variable is falsely issuing a recession probability above 50% for seven consecutive months due to a U.S. money market reform that pushed U.S. investors from prime money market funds into U.S. treasuries. Likewise, the FX basis seems to generate reasonable results within the framework of this analysis.

A closer look at the fundamental drivers, however, reveals three, conceptually different periods of basis widening: At the height of the GFC, the breakdown of the covered interest rate parity was primarily driven by interbank market stress and USD shortage (Borio et al. (2016)). The second period from 2011-2014 can be characterized by increasing redenomination risks in the Eurozone and the corresponding ex-ante excess demand for USD vs. the EUR. While the beginnings of both periods were associated with the start of the two recessions, the subsequent third period was merely a reflection of ECB excess liquidity following the accelerated bond purchase program. Hence, the latest widening has erroneously sent out recession signals for most of 2016. Hence, we suggest the use of implied rates volatility as a more robust input factor over the Libor-OIS spread and the EUR/USD FX basis which are more immediate but also more prone to exogenous shocks that are unrelated to the business cycle. The category *Preference for Risk* performs clearly better for risk-averse investors due to the higher sensitivity than for the risk-affine investor type who perceives missing a crisis as less costly. 19

The best performing variables in the *industrial data section* are retail sales (6m change) [see Figure 1g and Table A1 in the Appendix], stocks of finished products (6m change) and

<sup>17</sup> The cross-currency basis is the basis spread added to the USD LIBOR to re-establish the covered interest rate parity between the EUR and the USD.

<sup>&</sup>lt;sup>16</sup> The indicator dips briefly below the 50% threshold during those 14m.

<sup>&</sup>lt;sup>18</sup>Main drivers of the EUR/USD FX Basis are identified as reverse Yankee bond issuance and EU investors looking for higher yields in the U.S. (Borio et al. (2016)).

<sup>&</sup>lt;sup>19</sup> Among the total of all 110 transformations tested, this category (level specification) ranks 38<sup>th</sup> for risk-averse investors and only 55<sup>th</sup> for risk-affine investors.

new cars (12m change) with an AUROC between 0.78-0.83 and detected crisis between 45% and 47%. Given their coinciding nature with the business cycle, these variables serve to confirm recessions rather than being leading indicators.

Among inflation measures, a positive change of the inflation rate is surprisingly associated with an increase of the recession probability. An explanation for the leading characteristic of inflation could be its ability to indicate an overheating of the economy which, in turn, could trigger monetary tightening by the central bank. Although results for the 12m change of the HICP are relatively strong with an AUROC of 0.82, the variable is vulnerable to policy changes of the inflation target and structurally subdued inflation dynamics [see Table A1 in the Appendix].

Within the *commodities* section, findings reveal that an increase in the oil price is associated with an increasing likelihood of recession [see Figure 1g and Table A1 in the Appendix]. Although the oil price provides strong summary statistics with an AUROC of 0.82 and 62% of all crisis detected, the recent two years have proven that idiosyncratic supply side dynamics can severely taint the variable's informational content on the endogenous business cycle. For risk-averse investors, the variable ranks 6<sup>th</sup> while risk-affine investors find less utility placing the input factor 41st given the highest number of false recession signals among all variables (15%). This underlines the need to analyze underlying drivers of this variable despite sound overall statistical results.

Representing the wealth channel, both building permits (6m change) and the EuroStoxx50 (12m change) [see Figure 1h and Table A1 in the Appendix] do not qualify as reliable indicators given low AUROCs of 0.76 and 0.69 and only 40% and 16% of recession months detected. Building permits merely capture the first recession given the strong credit expansion in the run-up to the GFC and the related overheating of the housing market but fail to reflect the second economic downturn in the Eurozone. Detractors of the predictive power of stock market returns are numerous significant price corrections. As most of them were not followed by recessions, the probit model adjusts by showing a lower overall magnitude of swings in recession probability, crossing the 50% threshold only during the sell-off in 2002 and the GFC. Hence, risk-affine investors should not be advised to consider these variables as indicators for economic downturns given aforementioned fluctuations in valuations metrics.<sup>20</sup>

<sup>&</sup>lt;sup>20</sup> The average results for the variables reflecting the wealth channel rank 63<sup>rd</sup> for risk-affine investors and 47<sup>th</sup> for averse agents.

#### 6. Results of the Submodels

## 6.1 Submodel 1 – Monetary Transmission Channels

This submodel extracts the first principal component for categories representing different monetary transmission channels.<sup>21</sup> The idea behind this approach is not necessarily to outperform the results of the benchmark model, but rather finding a specification that removes idiosyncratic vulnerabilities of single indicators. We construct eight monetary transmission channels: Money supply competitiveness, sentiment, refinancing, preference for risk, industrial data, inflation and commodities.<sup>22</sup> Input factors are chosen based on a reasonable predictive quality in the previous univariate analysis. To ensure comparability among the different channels, we use all input factors in standardized form.<sup>23</sup>

The analysis yields the strongest improvements for following groups: Preference for risk, industrial data, and refinancing variables. All three categories are composed of variables that are sensitive to the economic cycle, but at the same time prone to idiosyncratic factors. By extracting the first principal component, it is possible to reduce these non-systemic fluctuations and extract the common driver of variability in the dataset.

#### < Insert Figure 2 >

Regarding preference for risk [see Figure 2a and Table 2A in the Appendix], the channel achieves an AUROC of 0.86 versus an average of 0.80 for its subcomponents. This finding is supported by a surge in accurately called recessions from an average of 47% to 62% and a fall in the number of false positive signals to 5% (average of 7%). Since both the Libor-OIS spread and the FX Basis have shown its vulnerability to exogenous shocks that do not stem from the economic cycle, the signals of the monetary transmission channels prove to be more robust.

The study finds similar results for the refinancing channel with an improvement of the AUROC by 6pp to 0.88 versus the average of its subcomponents which are idiosyncratically exposed to risks from European fragmentation and the HY market structure [see Figure 2b and Table 2A in the Appendix]. We find that particularly risk-averse investors benefit from this submodel as the percentage of crisis detected increases to 69% from 53%. This result underlines

<sup>&</sup>lt;sup>21</sup> For the detailed results of the submodels with respect to our employed information criteria see again the respective tables in the Appendix, while the subsequent figures only highlight the best performing indicators with respect to their probabilities of predicting a recession.

<sup>&</sup>lt;sup>22</sup> We exclude interest rates, the term spread and a potential wealth channel given the lack of adequate input factors.

<sup>&</sup>lt;sup>23</sup> Models consisting of more than two factors are established via principal component analysis, which requires standardization of the input factors.

the strength of the conceptual approach: While the HY and IG corporate bond spreads compensate for the failure of the Eurozone peripheral sovereign bond indicator during the first recession, the sustained positive recession signals of the latter in the run-up the second economic downturn is improving the channel's performance compared to the corporate spread variables.

Along the same lines, the channel of industrial variables indicates an improvement of the AUROC from an average of 0.80 to 0.89 of the combined model [see Figure 2c and Table 2A in the Appendix]. While new passenger cars and stocks of finished products do not serve as solid single indicators given a high number of false positive signals (average of 9%) and mediocre amount of crisis detected (average 46%), their combination with retails sales drives up the number of correctly predicted recessions to 65% and lowers the amount of false positive signals to 5%. Hence, even relatively volatile input factors can provide predictive value by their inclusion into existing models.

Regarding the competitiveness channel, we find a strong increase in accurately predicted recession months up to 71% from an average of 55% for the subcomponents [see Figure 2d and Table 2A in the Appendix]. The other monetary transmission channels (inflation, commodities, sentiment) yield results which are broadly in line with their underlying subcomponents. To provide insights into how much of the results is owed to the standardization of the variables, we included one model, *money supply*, which includes only one input factor. The results reveal that there is no material change as the AUROC improves marginally by 1pp [see Table A2 in the Appendix].

To conclude, the modeling of monetary transmission channels via principal component analysis yields significantly better results, in particular, for those channels whose subcomponents are most affected by idiosyncratic exogenous shocks. Besides statistical evidence, there is also a conceptual perspective that suggests higher future robustness associated with the use of aggregated channels even if results are in line with the underlying single input factors.

#### 6.2 Submodel 2 – Bivariate Specification

The second modification estimates bivariate probit regressions based on the six best single indicators from the benchmark model [see Figure 3 and Table A3 in the Appendix for the results]. To ensure that the selected variables are not close substitutes but complementary in nature, the paper first identifies the leading indicators of the different categories and then considers the six best-performing variables in all possible combinations. The chosen leading

indicators are: Real M1 (12m change), PMI (12m change), IG corporate bond spread (level), Terms of Trade (level), retail sales (6m change) and the oil price (level). In a second step, we conduct a similar exercise for the monetary transmission channels that cover the following six variables: Money supply, sentiment, refinancing, competitiveness, industrial data, and commodities.

The results of the bivariate probit models improve the AUROC on average by 0.06 pp to 0.93 vis-à-vis the average performance of the single underlying indicators. The stronger performance is reflected in a surge of crisis detected by 17pp to 0.76 and a reduction of false signals by 2pp to 0.09.

#### < Insert Figure 3 >

The supreme combination is Real M1 and the PMI which reaches a near-perfect AUROC of 0.99 with 95% correct signals and 91% of crisis detected (missing only five recession months) [see Figure 3a and Table A3 in the Appendix]. All other combinations that include Real M1 or the PMI perform marginally worse but still significantly better than the single input factors. Beneficial for the combination of Real M1 and the PMI is the leading nature of Real M1 in the run-up to the first recession, whereas the PMI remains elevated somewhat longer during the recession. This combination of complementary characteristics enables the bivariate model to improve the lead-time and cover the length of the economic downturn more accurately. Besides the paramount performance of bivariate models that include either Real M1 or the PMI, the combination of IG corporate bond spreads and crude oil leads the ranking of other potential pairs closely followed by the pair of Terms of Trade and retail sales.

Results for the monetary transmission channels underline the findings for the single input factors. As the concept of transmission channels introduces a greater variety of variables, we find this procedure to be balancing the superior statistical performance of bivariate analysis while reducing the inherent risk of overfitting.

#### 6.3 Submodel 3 – Unstable Surge

The third modification introduces the relative state of a variable versus its long-term trend in order to identify what Keynes termed "Animal Spirits" as well as potential hawkish mistakes by the central bank [see Table A4 in the Appendix for the results]. Traditional probit models tend to signal an elevated likelihood of a recession when indicators are deteriorating. However, the build-up of unsustainable momentum of economic developments can be an early indicator

of a looming correction following the possible collapse of over-optimism and inadequate production planning. Likewise, a hiking cycle which is too aggressive might be the harbinger of monetary tightening that has the potential to cause an economic downturn.

Hence, the idea is to detect periods of unstable surges measured against the Hodrick-Prescott (HP) trend estimated with a parameter value  $\lambda$  of 14,400 given the underlying monthly time series. The actual bivariate model consists of two input factors: 1) A single indicator that we conceptually suspect to describe these developments and 2) the deviation of this time series from its HP trend with three different lag structures (contemporaneous, with a 12m lag or a 24m lag). Potential indicators representing an overheating of the economy driven by animal spirits are: EuroStoxx50 (12m change), building permits (6m change), PMI (12m change), retail sales (6m change), MFI loans to the private sector (12m change) and crude oil (level). Input factors that are tested for their ability to detect a tightening of central bank policy are: Euribor (12m change), money supply (12m change) and inflation (12m change).

## < Insert Figure 4 >

Among the variables tested, results for the PMI (contemporaneous version) yield a surge in the number of recession months detected by 22pp up to 82% [see Figure 4a and Table A4 in the Appendix]. The lead-time towards the first recession rises from five months to 14 months while there is no improvement for the second recession as the economic downturn starting in 2011 was not preceded by an overly optimistic outlook of producers.

Notwithstanding the fact that growth of MFI loans does not detect a single recession month in the baseline model, the augmentation incorporating the HP deviation (contemporaneous version) yields an AUROC of 0.81 and correctly identifies 47% of all crisis periods [Figure 4b and Table A4 in the Appendix]. In addition, this transformation shows a remarkable lead-time of six months before the GFC and eight months before the second recession. Findings for retail sales (contemporaneous version) indicate particular appeal for risk-averse investors whose utility rises significantly [Figure 4c and Table A4 in the Appendix]. While the AUROC increases by 0.07 to 0.90, the number of detected crisis almost doubles from 45% to 78%. The flipside of these results is a higher number of false positive signals of 14% (up from 6%) which could be too noisy for risk-affine agents.

Results for building permits underline the strength and the incompleteness of this variable [Figure 4d and Table A4 in the Appendix]. Statistics point to a strong improvement of the AUROC to 0.89 up from 0.76. However, this is mainly due to a more accurate reflection of

animal spirits in the run-up to the first recession. Given the lack of causality, the augmented indicator also misses out completely on the second economic downturn. Results for the other indicators do not show improved test statistics as both EuroStoxx50 and crude oil remain volatile, while the indicator of retail sales does not recede in between the two recessions.

With regard to hawkish mistakes by the central bank, we find that the augmented variable of the 3m Euribor carries predictive value [Figure 4g and Table A4 in the Appendix]. While the indicator does not call any recession month in the univariate benchmark model, the modification including a one-year lag results in an AUROC of 0.86 and 65% of recession months detected. The augmentation captures both recessions with a lead-time of ten months (first recession) and six months (second recession), indicating that an above-trend increase of the Euribor might signal the central bank's attempt to tighten monetary supply. In particular, the two hikes in the run-up to the second recession represent a poster model for a hawkish mistake. On the inflation variable, results suggest that a positive deviation from trend increases the probability of a recession [Figure 4h and Table A4 in the Appendix]. As above-trend inflation could induce a central bank to tighten rates, this indicator yields a moderate increase of recessions detected to 62% (up from 53%). Results for money supply are broadly in line with the univariate benchmark analysis with an AUROC inching slightly higher to 0.97 (up from 0.95).

To conclude, the addition of the deviation of indicators from their HP trend yields significant improvements for some of the variables as they are potential witnesses to unsustainable surges. Nevertheless, the results need to be interpreted with some caution as the HP trend is quite sensitive to the latest values of a time series and the ex-post analysis does not replicate a real-time assessment.

#### 7. Out-of-Sample Analysis and Varying Time Horizons

#### 7.1 Out-of-Sample Analysis

The out-of-sample test is calibrated for the time period between December 2001 and August 2010. Hence, it covers the period leading up to the GFC, the full length of the first recession and 14 months of the subsequent expansion phase, leaving a buffer of 13 months until the start of the second recession. The detailed results for the benchmark model are displayed in Table A5 in the Appendix, while Tables A6 through A8 in the Appendix show the result of the out-of-sample analysis for the three submodels.

Regarding the univariate benchmark model, the analysis allows a broad classification into two groups of indicators: Those variables that are rather unspecific in nature and a

reflection of broader business cycle dynamics as well as those indicators that are specific to the respective recession. Indicators that display the ability to capture both recessions are Real M1, the PMI, retail sales, implied interest rate volatility, the Libor-OIS spread, and crude oil whose summary statistics are similar to those of the benchmark model from section 5. Variables falling short of predicting both economic downturns are the term spread, the Eurozone peripheral sovereign bond spread, inflation and building permits. In particular the Eurozone peripheral sovereign bond spread underlines the necessity to qualitatively assess the probability of potential structural breaks: While investors have assumed an implicit guarantee of Eurozone government debt by other members of the currency bloc before the European debt crisis, the indicator has become functional after markets have revised this assumption in light of redenomination risks. On the same note, the potential formation of a European Monetary Fund, as currently discussed by policymakers, could represent a development that decreases the predictive quality of the indicator given lower sensitivity of spread levels.

#### 7.2 The Benchmark Model over Various Time Horizons

To introduce an additional dimension, the analysis modifies the forecasting horizon to 6m, 18m and 24m to shed further light on the leading, coinciding or lagging nature of potential indicators.

For the 6m forecasting horizon, both industrial data and sentiment indicators gain substantially in predictive quality [see Table A9 in the Appendix for the details]. The PMI (12m change) becomes the best performing indicator with an AUROC of 0.95 (+0.07)<sup>24</sup> and 74% of crisis detected (+14pp). This result is reflective for the broader group of sentiment variables whose AUROCs increase to an average of 0.86 (up from 0.73). With regard to the category of industrial variables, we find very strong performance of retail sales (6m change) and stocks of finished products (12m change) whose AUROCs increase to 0.90 (+0.07) and 0.89 (+0.12) respectively on the back of better crisis detection and less false positive signals. Findings for both categories emphasize their particular usefulness in validating recessions given their shorter lead-time. Both buildings permits and the EuroSTOXX50 improve their results for the 6m horizon but still clearly underperform other indicators. For the other categories, we find a tendency of fewer crises detected and less false signals. As a consequence, models covering shorter time horizons provide a valuable perspective for risk-affine investors who face higher costs of false positive signals.

For the 18m horizon [see Table A10 in the Appendix], sentiment indicators and industrial data deteriorate across the board and lose out on their predictive capabilities, while

<sup>&</sup>lt;sup>24</sup> The number in brackets indicates the relative change to the results of the 12m forecasting horizon.

financial market spread indicators closely track Real M1 as the best performing variables due to their longer-lead time. As to the forecasting ability of Real M1, the percentage of crisis detected remains practically unchanged at 75% versus the benchmark model, but false positive signals increase by 10pp to 15%. A common observation for the 18m forecasting horizon is a higher number of false positive signals which raises particularly the costs for the risk-affine investor type.

Considering the 24m forecasting horizon, findings indicate that it is hardly possible to achieve reliable forecasting performance [see Table A11 in the Appendix]. In particular, the number of false alarm surges across the dataset, declaring on average 18% of all expansion months to recessions (versus 3% for the 12 month forecasting horizon). Only two variables stand out with a slight deterioration against the 18m horizon: Both the peripheral sovereign spread and the IG corporate bond spread yield only moderately weaker results and could serve risk-averse investors as early warning indicators with a considerable lead-time.

#### 7.3 Submodels over Various Time Horizons

Considering the first submodel, results indicate superior performance of the monetary transmission channels incorporating industrial data, sentiment variables and money supply with AUROCs between 0.95 and 0.93 and 72% to 70% of recession months correctly called<sup>25</sup> [see Table A12 in the Appendix for all time horizons]. Looking at the extended forecasting periods of 18 to 24 months, we note for sentiment variables an expected decrease in their predictive abilities (AUROC of 0.69) whereas refinancing variables perform reasonably well with an AUROC of 0.83 and 66% of recessions detected. Surprisingly good predictive qualities for the longer term horizon are associated with the channel capturing the preference for risk. In particular risk-affine investors will appreciate only 12% false positive signals (versus 23% of the other channels).

Based on bivariate specifications, the second submodel continues to yield significantly better results than their underlying univariate time series [Table A13 in the Appendix]. With an average AUROC of 0.98 and 82% of crisis months detected across all possible combinations, the PMI and Real M1 are the preferred indicators for the 6m horizon. In light of their shorter lead-time, channels composed of industrial data and sentiment variables display very strong predictive ability. Lengthening the time horizon to 18 months reveals a substantial increase in

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<sup>&</sup>lt;sup>25</sup> While preference for risk and the competitiveness channel yield overall strong results, they are lagging either with regard to the number of recession detected (60% for preference of risk) or a higher number of false alarm (8% for competitiveness)

the number of false positive signals, whereas monetary transmission channels perform better than their univariate peers with 11% false alarm versus 14%.<sup>26</sup> A reason for the higher stability of monetary transmission channels could be the inclusion of variables that are not perfectly congruent in their lead-time.

Regarding the third submodel of unstable surges, the analysis reveals for the 6 months horizon an AUROC of 1 for Real M1 with 93% of all recessions months correctly identified [see Table A14 in the Appendix]. Even for the 18m horizons, this augmentation detects 79% of all crises while issuing 15% false positive signals. Both retail sales and MFI loans prove robust for the shorter term horizon with an AUROC of 0.94 (0.90 in the baseline model) and 0.82 respectively (0.81 in the baseline model). The further we move out on the forecasting horizon, the more important becomes the calibration of the lag structure. In light of the results for the 18m horizon, we suggest a coinciding assessment of the cyclical stance (contemporaneous lag) as compared to a lag of 12 months or 24 months for the shorter-term forecasting horizons. Contrary to previous results that render retails sales rather useless for the longer term perspective, the third submodel yields an AUROC of 0.88 and 80% of crisis detected for retail sales over the 24 month horizon.

#### 8. Summary of Results

In order to assess recession probabilities for the Eurozone, this paper applied four different approaches: A univariate benchmark model to identify predictors and their preferred transformations as well as three submodels. For all four approaches out-of-sample analyses are performed as well. To increase robustness and to capture additional dynamics, various forecast horizons next to the benchmark of 12 months were employed to validate the lead-time. In order to measure the predictive power of the indicator variables we rely on three different signaling approaches to come up with robust results: A signaling approach (Kaminsky and Reinhart (1999)), the area under the receiver operating curve (AUROC) and a loss function to model risk-averse and risk-seeking investors (Alessi and Detken (2009)).

The regime change of monetary policy at the zero lower bound is posing challenges to the empirical supremacy of the term spread for signaling recessions. We find that a modified version of the yield curve incorporating the shadow rate by Xia and Wu (2016) removes the downward rigidity of the front-leg and restores part of the informational content of the yield curve. However, regarding the 12m forecasting horizon, the best performing single indicator of the analysis is Real M1 with a lead-time of eleven months before the start of both recessions

<sup>&</sup>lt;sup>26</sup> The average number of crisis detected is similar with 73% for channels versus 72% for univariate indicators.

under consideration. Slightly trailing in performance, the Purchasing Managers Index, the investment grade corporate bond spread and the Terms of Trade yield the highest predictive quality. Considering the 12m transformation of variables instead of their level increases the sensitivity to turning points of the business cycle for those variables that are less leading in nature. Other variables (e.g., Libor-OIS, crude oil) generate good summary statistics but reveal idiosyncratic vulnerabilities that can lead to a higher number of false positive signals.

For risk-averse investors, who are more concerned with not missing a crisis at the expense of a higher number of false positive signals, there is higher relative utility in considering the price of crude oil and Terms of Trade. Results for risk-affine investors suggest a higher relative reliance on variables that are producing fewer false alarms such as HICP inflation and the Eurozone peripheral sovereign bond spread.

On the first submodel, the creation of monetary transmission channels yields the strongest improvements for the categories of industrial data, refinancing variables and preference for risk. As all those three categories are composed of variables prone to idiosyncratic fluctuations, the use of principal component analysis increases the informational value regarding the business cycle. Results for the second submodel suggest bivariate combinations perform significantly better than the univariate specification. In particular, combinations of indicators including Real M1 or the PMI can be considered to be paramount. Bivariate models of monetary transmission channels also outperform their univariate equivalents by far and, in addition, reduce the risk of overfitting compared to bivariate models of single indicators. Considering the deviation from the HP trend in the third submodel achieves significant improvements for indicators in the run-up to the recession of 2008-2009 which can be characterized by signs of overheating. Similarly, monetary policy variables become more sensitive through suggested modification by capturing potential hawkish mistakes.

Regarding different forecasting horizons, sentiment indicators and industrial data improve significantly for shorter time horizons, whereas corporate spreads and the Eurozone peripheral sovereign bond spread underline their leading character vis-à-vis the business cycle given a remarkable performance over longer forecasting horizon. Most of the indicators fail to be useful over the 24m horizon given a high number of false signals. However, indicators of the third submodel reflecting unstable surges provide moderately robust results also for extended forecasting horizons that could serve as complements to short-term forecasts.

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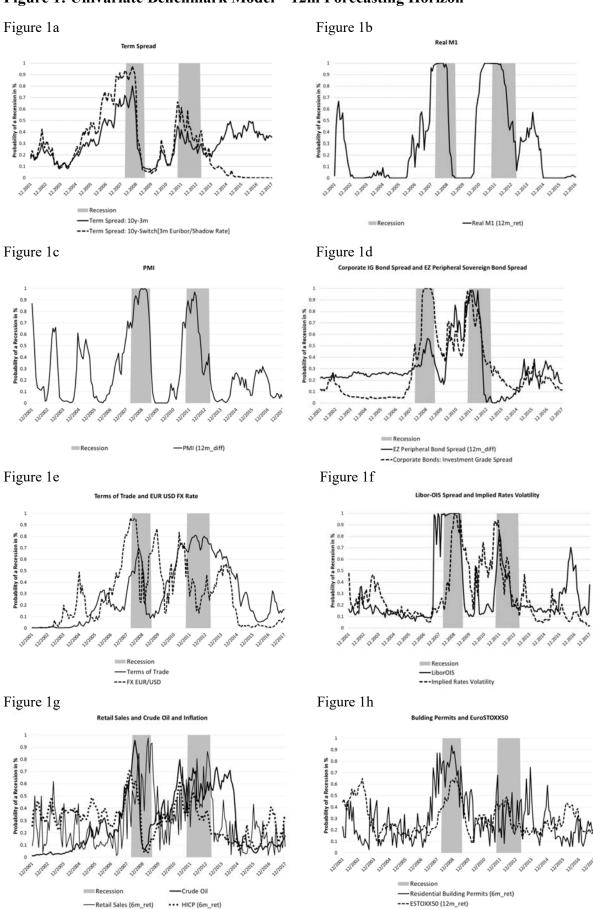
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Figure 1: Univariate Benchmark Model – 12m Forecasting Horizon



**Figure 2: Submodel 1 - Monetary Transmission Channels** 



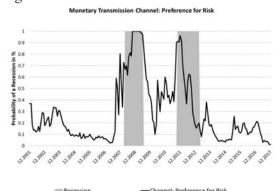


Figure 2b

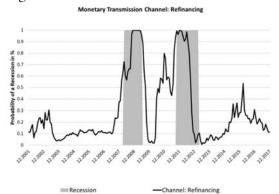


Figure 2c

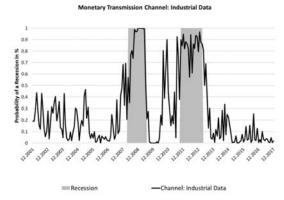


Figure 2d



Figure 2e

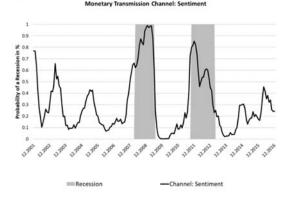


Figure 2f

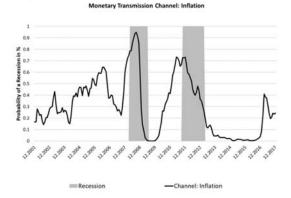


Figure 2g

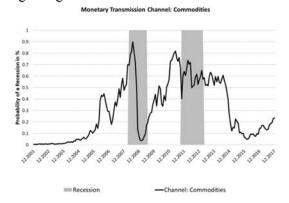


Figure 2h

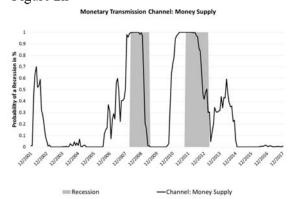


Figure 3: Submodel 2 - Bivariate Specification



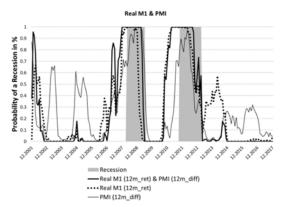


Figure 3c

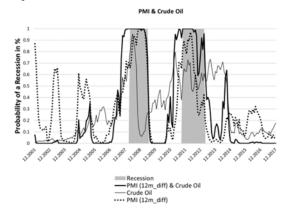


Figure 3e

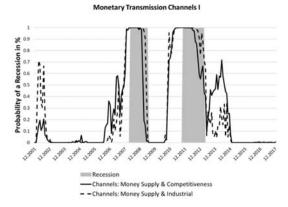


Figure 3g

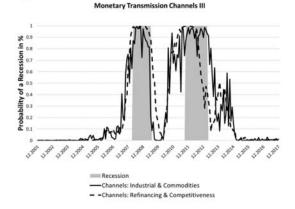


Figure 3b

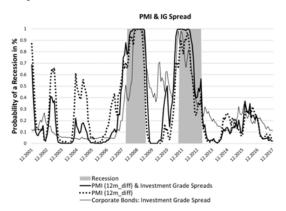


Figure 3d

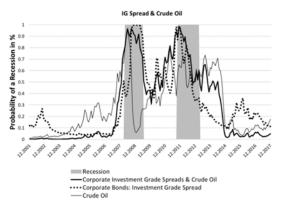


Figure 3f

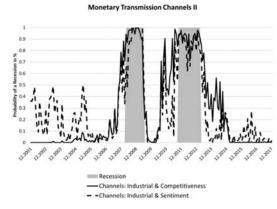


Figure 3h

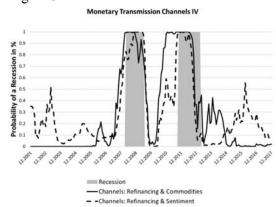


Figure 4: Submodel 3 - Unstable Surges



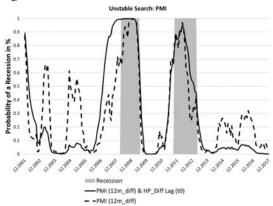


Figure 4c

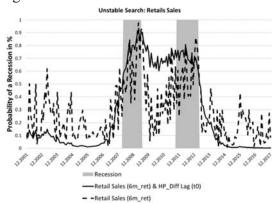


Figure 4e

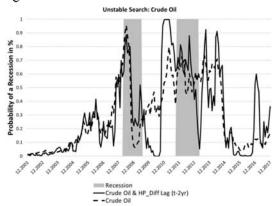


Figure 4g

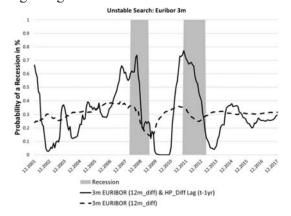


Figure 4b

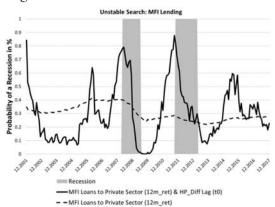


Figure 4d

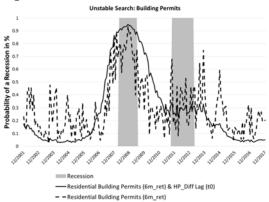


Figure 4f

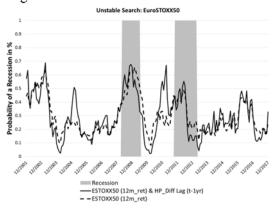
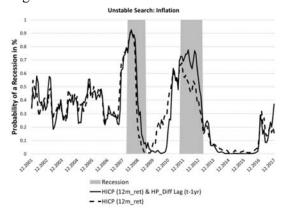


Figure 4h



**Table 1: List of Variables** 

Variables		C	C. I.		
variables		Source	Code	Transformation	Comment
Rates	3m EURIBOR	Haver	N023RI3E@G10	***	
	10yr German Government Bond	BBG	GTDEM10Y Govt	***	
Term Spread	Term Spread: 10y-3m	BBG / Haver	10yr: GTDEM10Y /3m: N023RI3E@G10	***	
	Term Spread: 10y-Switch[3m Euribor/Shadow Rate]	Haver	N023SHDW@G10	***	
	M1	Haver	S023FM1@G10	**	
	Real M1	Haver	Inflation: H025PCXG@G10	**	
Money Supply	M2	Haver	S023FM2@G10	**	
one y suppry	Real M2	Haver	Inflation: H025PCXG@G10	**	
	M3	Haver	S023FM3@G10	**	
	Real M3	Haver	Inflation: H025PCXG@G10	**	
	FX EUR/USD	Haver	N023XUNE@G10	***	
Competitiveness	JPM Nominal Broad Effective FX Rate	Haver	N023XJNB@G10	***	
	Terms of Trade	Haver	H025PFTT@G10	***	Time lag of 2 months
	PMI	Haver	M023TG@MKTPMI	***	
	Economic Sentiment	Haver	S025VES@G10	***	
Sentiment	Industrial Confidence	Haver	S025VB@G10	***	
Sentiment	Consumer Confidence	Haver	S025VCC@G10	***	
	Consumer Economic Situation	Haver	S025VCE@G10	***	
	Business Climate	Haver	S025VBC@G10	***	
	Corporate Bonds: Investment Grade Spread	Barclays	2002	***	Option-adjusted spread
	High Yield Spread	ICE	HPC4	***	Option-adjusted spread
Refinancing	EZ Peripheral Bond Spread	BBG	BBG 10yr generic yields of DE/IT/ES/GR/IE/PT	***	Country weighting of 2017 = 2016 given GDP publication lag; Greece is included since 2009
	MFI Loans to Private Sector	Haver	S023FCB@G10	**	
	Implied Rates Volatility	BBG	EUNEF10 ICPL/ EUSN0F10 / EUSV0F10	***	
	VSTOXX	BBG	V2X Index	***	
Dist. Assessing	VIX	BBG	VIX Index	***	
Risk Aversion	FX Basis EUR/USD	BBG	EUBS1 Index	***	
	Implied FX Vol	BBG	EURUSDV1M Index	***	
	LiborOIS	BBG	BICLOISS Index	*	
	Industrial Production: Industry excl. Constr.	Haver	S025D@G10	**	Time lag of 1 month
	Unemployment Rate	Haver	S023ELUR@G10	***	Time lag of 1 month
	Manufact. New Orders	Haver	S025TO@G10	**	Time lag of 1 month
Industrial Data	Industr. Turnover: Manufacturing	Haver	S025TS@G10	**	Time lag of 1 month
	MfG Stocks of Finished Products	Haver	S025TI@G10	***	
	Retail Sales	Haver	S025TRSC@G10	**	Time lag of 1 month
	New Passenger Cars	Haver	H025CVRT@G10	**	
	HICP	Haver	H023PC@G10	**	
Inflation	HICP excl. Energy, Food, Alcohol & Tobacco	Haver	H025PCXG@G10	**	
	Domestic PPI: Industry excluding Construction	Haver	H025PP@G10	**	Time lag of 1 month
	Crude Oil	BBG	CL1 Comdty	***	-
Commodities	Gold	BBG	XAU Comdty	***	
	ESTOXX50	Haver	N023FKAE@G10	**	
Wealth	Residential Buidling Permits	Haver	S025HP@G10	**	Time lag of 1 month
			1		

<sup>\*</sup> Level; \*\* 6m and 12m change; \*\*\* Level, 6m and 12m change

# Appendix

**Table A1: Univariate Benchmark Model – 12m Forecasting Horizon** 

12m Forecas	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification Signal Issued	Recession	No Recession	True positive vs.	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	L = f * (Type2Error)	+ (1 - f) * (Type1Erro
Channel	No Signal Issued 3m EURIBOR	A C	B D	False Positive Rate	80%	33%	0%	L= 0.25 0.17	L= 0.75 0.50
	3m EURIBOR (6m_diff)			46%	72%	9%	0%	0.23	0.68
Rates	3m EURIBOR (12m_dift 10yr German Governm			63% 53%	70% 70%	0% 0%	0%	0.25	0.75 0.75
	10yr German Governm	ent Bond (6m_diff)		52%	70%	0%	0%	0.25	0.75 0.75
	10yr German Governm Term Spread: 10y-3m	ent Bond (12m_diff)		53% 69%	70% 78%	0% 33%	0% 2%	0.25	0.75
	Term Spread: 10y-3m ( Term Spread: 10y-3m (			53% 66%	70% 70%	0% 0%	0% 0%	0.25	0.75 0.75
Term Spread	Term Spread: 10y-Swit	ch[3m Euribor/Shadow R	ate]	84%	80%	49%	7%	0.18	0.40
	Term Spread: 10y-Swit Term Spread: 10y-Swit	ch[3m Euribor/Shadow R ch[3m Euribor/Shadow R	ate] (6m_diff) atel (12m_diff)	57% 74%	70% 72%	0% 16%	0% 4%		0.75 0.64
	M1 (6m_ret)		,	84%	84%	62%	6%	0.14	0.30
	M1 (12m_ret) Real M1 (6m_ret)			94% 85%	86% 86%	69% 65%	7% 5%	0.13	0.25
	Real M1 (12m_ret) M2 (6m_ret)			95% 52%	88% 70%	73% 0%	5% 0%	0.10	0.22 0.75
Money Supply	M2 (12m ret)			53%	70%	0%	0%	0.25	0.75
Worley Supply	Real M2 (6m_ret) Real M2 (12m_ret)			49% 50%	70% 70%	0%	0%		0.75 0.75
	M3 (6m_ret)			51%	70%	0%	0%	0.25 0.25 0.18 0.26 0.24 0.14 0.13 0.12 0.10 0.25 0.25 0.25 0.25 0.25 0.25 0.25 0.2	0.75
	M3 (12m_ret) Real M3 (6m_ret)			53% 49%	70% 70%	0% 0%	0% 0%		0.75 0.75
	Real M3 (12m_ret) FX EUR/USD			51% 84%	70% 78%	0% 49%	0% 10%		0.75 0.41
	FX EUR/USD (6m_ret)			48%	70%	0%	0%	0.25	0.75
	FX EUR/USD (12m_ret) JPM Nominal Broad Ef	fective FX Rate		53% 67%	70% 70%	0% 16%	0% 6%		0.75 0.64
Competitiveness	JPM Nominal Broad Ef	fective FX Rate (6m_ret)		53%	70%	0%	0%	0.25	0.75
	JPM Nominal Broad Ef Terms of Trade	fective FX Rate (12m_ret		61% 85%	69% 78%	0% 60%	2% 13%		0.75 0.33
	Terms of Trade (6m_re	t)		63%	70%	0%	0%	0.25	0.75
	Terms of Trade (12m_r PMI	erj		70% 69%	70% 79%	22% 40%	10% 4%	0.18	0.61
	PMI (6m_diff)			80% 88%	81% 83%	51% 60%	6% 7%		0.38
	PMI (12m_diff) Economic Sentiment			57%	72%	13%	2%	0.23	0.32 0.66
	Economic Sentiment (6 Economic Sentiment (1			82% 76%	82% 81%	56% 47%	6% 5%		0.34
	Industrial Confidence			48%	70%	0%	0%	0.25	0.75
	Industrial Confidence ( Industrial Confidence (			81% 74%	82% 74%	51% 27%	5% 6%		0.38
Sentiment	Consumer Confidence			69%	75%	27%	5%	0.22	0.56
	Consumer Confidence Consumer Confidence			77% 73%	81% 78%	51% 49%	6% 10%		0.38
	Consumer Economic Si Consumer Economic Si			75% 76%	83% 80%	55% 47%	5% 6%		0.35 0.41
	Consumer Economic Si			79%	80%	51%	8%	0.18	0.39
	Business Climate Business Climate (6m_	diff)		48% 81%	70% 82%	0% 51%	0% 5%		0.75 0.38
	Business Climate (12m	_diff)		75%	73%	25%	6%	0.23	0.57
	Corporate Bonds: Inve Corporate Bonds: Inve	stment Grade Spread stment Grade Spread (6n	n diff)	85% 61%	82% 77%	56% 24%	7% 0%		0.35
	Corporate Bonds: Inve	stment Grade Spread (12		78%	82%	44%	1%	0.15	0.42
	High Yield Spread High Yield Spread (6m	_diff)		76% 65%	71% 77%	25% 31%	9% 2%	0.19	0.58 0.52
Refinancing	High Yield Spread (12n EZ Peripheral Bond Spr	n_diff)		76% 74%	85% 80%	55% 45%	2% 6%		0.35 0.42
	EZ Peripheral Bond Spi	ead (6m_diff)		77%	71%	11%	2%	0.24	0.67
	EZ Peripheral Bond Spr MFI Loans to Private Se			85% 52%	82% 70%	49% 0%	3% 0%		0.39 0.75
	MFI Loans to Private Se	ector (12m_ret)		58%	70%	0%	0%	0.25	0.75
	Implied Rates Volatilit Implied Rates Volatilit	(6m_diff)		79% 67%	80% 73%	51% 20%	8% 4%		0.39
	Implied Rates Volatility VSTOXX	/ (12m_diff)		71% 66%	75% 69%	24% 13%	3% 6%		0.58 0.67
	VSTOXX (6m_ret)			56%	71%	5%	1%	0.24	0.71
	VSTOXX (12m_ret) VIX			65% 73%	71% 71%	13% 22%	4% 8%		0.66 0.61
Risk Aversion	VIX (6m_ret)			58%	71%	11%	2%	0.24	0.67
	VIX (12m_ret) FX Basis EUR/USD			70% 78%	79% 75%	40% 38%	4% 9%		0.46 0.49
	FX Basis EUR/USD (6mg FX Basis EUR/USD (12mg			53% 69%	70% 76%	0% 20%	0% 0%	0.25	0.75 0.60
	Implied FX Vol			64%	76%	24%	1%	0.20	0.57
	Implied FX Vol (6m_ret Implied FX Vol (12m_ret			57% 64%	70% 71%	7% 15%	2% 5%		0.70 0.65
	LiborOIS		rot	82%	82%	53%	6%	0.16	0.37
	Industrial Production:	Industry excl. Constr. (6n Industry excl. Constr. (12		65% 52%	73% 70%	15% 0%	2% 0%	0.25	0.64
	Unemployment Rate Unemployment Rate (6			58% 65%	70% 71%	0% 9%	0% 2%	0.25	0.75 0.69
	Unemployment Rate (	L2m_ret)		56%	70%	0%	0%	0.25	0.75
	Manufact. New Orders Manufact. New Orders			65% 55%	75% 70%	18% 0%	1% 0%		0.62 0.75
Industrial Data	Industr. Turnover: Mar	nufacturing (6m_ret)		54%	72%	9%	1%	0.23	0.68
Dayarur Data	Industr. Turnover: Mar MfG Stocks of Finished			60% 48%	70% 70%	0% 0%	0% 0%	0.25 0.25	0.75 0.75
	MfG Stocks of Finished MfG Stocks of Finished	Products (6m_diff)		79% 77%	78% 78%	47% 42%	8% 6%	0.19	0.42 0.45
	Retail Sales (6m_ret)	1 Toducts (12III_GIII)		83%	80%	45%	6%	0.18	0.42
	Retail Sales (12m_ret) New Passenger Cars (6	m ret)		78% 69%	76% 71%	44% 13%	10% 3%	0.21 0.24	0.45 0.66
	New Passenger Cars (1			78%	77%	45%	10%	0.21	0.43
	HICP (6m_ret) HICP (12m_ret)			76% 82%	77% 83%	25% 53%	0% 3%	0.19 0.14	0.56 0.36
Inflation	HICP excl. Energy, Food	I, Alcohol & Tobacco (6m		68%	65%	2%	7%	0.30	0.75
	Domestic PPI: Industry	I, Alcohol & Tobacco (12 excluding Construction (	m_ret) 6m_ret)	69% 72%	63% 74%	0% 15%	10%	0.32	0.77
	Domestic PPI: Industry	excluding Construction (	12m_ret)	80%	76%	51%	13%	0.22	0.40
	Crude Oil (6m_diff)			82% 47%	78% 70%	62% 0%	15% 0%	0.21 0.25	0.32 0.75
Commodities	Crude Oil (12m_diff) Gold			52% 76%	70% 82%	0% 49%	0% 3%	0.25 0.15	0.75 0.39
	Gold (6m_diff)			58%	70%	2%	1%	0.25	0.74
	Gold (12m_diff) ESTOXX50 (6m_ret)			60% 66%	69% 71%	2% 18%	2% 6%	0.26 0.25	0.74 0.63
Wealth	ESTOXX50 (12m ret)			69%	68%	16%	10%	0.28	0.65
	Residential Building Pe	:4- (C+)		76%	77%	40%	6%	0.20	0.47

<sup>&</sup>quot;diff" describes the absolute difference between levels; "ret" describes the % change between levels;

**Table A2: Submodel 1 - Monetary Transmission Channels** 

12m Forecas	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive vs.				L = f * (Type21	Error) + (1 - f) *
Channel	Signal Issued	Α	В	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		
	No Signal Issued	С	D					L= 0.25	L= 0.75
Channel: Money				96%	90%	76%	5%	0.09	0.19
Real M1 (12r				95%	88%	73%	5%	0.10	0.22
Channel: Compe	itiveness			89%	80%	71%	16%	0.19	0.26
FX EUR/USD				84%	78%	49%	10%	0.20	0.41
Terms of Tra				85%	78%	60%	13%	0.20	0.33
Channel: Sentime				80%	83%	56%	6%	0.15	0.34
PMI (12m_di				88%	83%	60%	7%	0.15	0.32
	ntiment (6m_diff)			82%	82%	56%	6%	0.16	0.34
	ntidence (6m_ditt)			81%	82%	51%	5%	0.16	0.38
	nate (6m_diff)			81%	82%	51%	5%	0.16	0.38
	onomic Situation (12m_	_diff)		79%	80%	51%	8%	0.18	0.39
	onfidence (6m_diff)			77%	81%	51%	6%	0.16	0.38
Channel: Refinan				88%	87%	69%	5%	0.11	0.24
Corporate Bo	nds: Investment Grade	Spread		85%	82%	56%	7%	0.16	0.35
EZ Periphera	Bond Spread (12m_diff	1)		85%	82%	49%	3%	0.15	0.39
High Yield Sr	read (12m diff)			76%	85%	55%	2%	0.13	0.35
Channel: Prefere	nce for Risk			86%	85%	62%	5%	0.13	0.30
LiborOIS				82%	82%	53%	6%	0.16	0.37
Implied Rate	s Volatility			79%	80%	51%	8%	0.18	0.39
FX Basis EUR	/USD			78%	75%	38%	9%	0.22	0.49
Channel: Industr	al Data			89%	86%	65%	5%	0.12	0.27
Retail Sales (	6m_ret)			83%	80%	45%	6%	0.18	0.42
MfG Stocks o	f Finished Products (6m	diff)		79%	78%	47%	8%	0.19	0.42
	er Cars (12m_ret)			78%	77%	45%	10%	0.21	0.43
Channel: Inflatio				84%	80%	53%	8%	0.18	0.37
HICP (12m r	et)			82%	83%	53%	3%	0.14	0.36
	: Industry excluding Con	struction (12m ret)		80%	76%	51%	13%	0.22	0.40
Channel: Commo				85%	80%	71%	16%	0.19	0.26
Crude Oil				82%	78%	62%	15%	0.21	0.32
Gold				76%	82%	49%	3%	0.15	0.39
55.5				7070	0270	4570	3/0	0.15	0.55

**Table A3: Submodel 2 - Bivariate Specification** 

12m Forecas	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive vs.				L = f * (Type2I	Error) + (1 - f) *
Channel	Signal Issued No Signal Issued	A C	B D	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	L= 0.25	L= 0.75
	Real M1 (12m_ret) & Pl			99%	95%	91%	3%	0.05	0.08
		vestment Grade Spread:	3	97% 96%	93%	82%	2%	0.06	0.14
		Real M1 (12m_ret) & Terms of Trade			91%	78%	4%	0.08	0.17
	Real M1 (12m_ret) & Re			97%	93%	89%	5%	0.06	0.09
	Real M1 (12m_ret) & Crude Oil			96% 93%	91%	76%	2%	0.08	0.18
		MI (12m_diff) & Investment Grade Spreads			87%	64%	3%	0.11	0.28
	PMI (12m_diff) & Term:			97%	91%	78%	3%	0.08	0.17
Single Indicators	PMI (12m_diff) & Retail			90%	85%	71%	9%	0.14	0.24
	PMI (12m_diff) & Crude			98% 90%	92%	82%	4%	0.08	0.15
		orporate Investment Grade Spreads & Terms of Trade orporate Investment Grade Spreads & Retail Sales (6m ret)			86%	78%	11%	0.14	0.19
				87%	83%	62%	7%	0.15	0.30
		Grade Spreads & Crude (	Dil	93%	86%	76%	10%	0.13	0.20
	Terms of Trade & Retail Sales (6m_ret)			90%	86%	78%	10%	0.13	0.19
	Terms of Trade & Crude Oil			86% 89%	79%	65%	15%	0.20	0.30
		Retail Sales (6m_ret) & Crude Oil			84%	75%	12%	0.15	0.22
	Channels: Money Supp			98%	93%	87%	4%	0.06	0.11
	Channels: Money Supp			98%	94%	87%	3%	0.06	0.10
	Channels: Money Supp			97%	91%	82%	5%	0.08	0.15
	Channels: Money Supp			98%	94%	87%	3%	0.06	0.10
	Channels: Money Supp			97%	92%	82%	3%	0.07	0.14
	Channels: Industrial & I			93%	88%	73%	6%	0.11	0.22
	Channels: Industrial &			96%	90%	82%	7%	0.10	0.15
Channels	Channels: Industrial & 5			90%	87%	69%	5%	0.11	0.24
	Channels: Industrial &			96%	91%	87%	7%	0.09	0.11
	Channels: Refinancing			96%	91%	78%	4%	0.08	0.17
	Channels: Refinancing			89%	86%	62%	3%	0.12	0.29
	Channels: Refinancing			97% 97%	92%	82%	3%	0.07	0.14
		Channels: Competitiveness & Sentiment			91%	78%	4%	0.08	0.17
	Channels: Competitive			90%	81%	73%	16%	0.19	0.24
	Channels: Sentiment &	Commodities		98%	91%	84%	6%	0.09	0.14

**Table A4: Submodel 3 - Unstable Surges** 

	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification Sianal Issued	Recession A	No Recession B	True positive vs.	(A+D)/(A+B+C+D)	A/(A+C)	A / (B+D)	L = f * (Type2Error) -	(1 - f) * (Type1Error
	No Signal Issued	c	D	False Positive Rate	, ,,,	, , , ,	, , ,	L= 0.25	L= 0.75
	Real M1 & HP_Diff Lag			97%	90%	80%	6%	0.09	0.16
	Real M1 & HP_Diff Lag	(t-1yr)		97%	92%	84%	5%	0.08	0.13
Money Supply	Real M1 & HP_Diff Lag			96%	90%	78%	5%	0.09	0.18
	Single Indicator: Real M1 (12m_ret)			95%	88%	73%	5%	0.10	0.22
	ESTOXX50 (12m_ret) &			68%	70%	18%	7%	0.26	0.63
Carrier Manufacta	ESTOXX50 (12m_ret) &			71%	75%	33%	7%	0.22	0.52
Equity Markets	ESTOXX50 (12m ret) & HP Diff Lag (t-2yr)			69%	69%	16%	9%	0.27	0.65
	ESTOXX50 (12m ret)			69%	68%	16%	10%	0.28	0.65
		ermits (6m_ret) & HP_Di		89%	76%	45%	11%	0.22	0.44
D.::141		ermits (6m_ret) & HP_Diff		81%	81%	53%	6%	0.17	0.37
Building Permits	Residential Building Pe	ermits (6m_ret) & HP_Diff	Lag (t-2yr)	76%	77%	40%	6%	0.20	0.47
		ng Permits (6m_ret)		76%	77%	40%	6%	0.20	0.47
	PMI (12m_diff) & HP_I			94%	91%	82%	6%	0.09	0.15
	PMI (12m_diff) & HP_D			90%	81%	62%	11%	0.18	0.31
Sentiment	PMI (12m_diff) & HP_D			90%	82%	64%	10%	0.17	0.30
	Single Indicator: P			88%	83%	60%	7%	0.15	0.32
	HICP (12m_ret) & HP_D			82%	85%	55%	2%	0.13	0.35
Inflation	HICP (12m_ret) & HP_I			84%	85%	62%	5%	0.13	0.30
inflation	HICP (12m_ret) & HP_D	Diff Lag (t-2yr)		84%	82%	56%	6%	0.16	0.34
	Single Indicator: H			82%	83%	53%	3%	0.14	0.36
	Retail Sales (6m_ret) 8			90%	83%	78%	14%	0.16	0.20
Retail	Retail Sales (6m_ret) &	HP_Diff Lag (t-1yr)		85%	79%	51%	9%	0.19	0.39
Ketali	Retail Sales (6m_ret) &			83%	79%	44%	6%	0.18	0.44
		letail Sales (6m_ret)		83%	80%	45%	6%	0.18	0.42
		ector (12m_ret) & HP_Di		81%	78%	47%	9%	0.20	0.42
Credit Dynamics		ector (12m_ret) & HP_Dif		70%	70%	4%	1%	0.25	0.72
Credit Dynamics		ector (12m_ret) & HP_Dif		62%	72%	9%	1%	0.23	0.68
		AFI Loans to Private Secto	r (12m_ret)	58%	70%	0%	0%	0.25	0.75
	Crude Oil & HP_Diff La			88%	76%	53%	13%	0.22	0.39
Oil	Crude Oil & HP_Diff Lag			84%	76%	56%	16%	0.23	0.37
Oil	Crude Oil & HP_Diff Lag (t-2yr)			90%	85%	71%	9%	0.14	0.24
	Single Indicator: C			82%	78%	62%	15%	0.21	0.32
	3m EURIBOR (12m_diff			63%	70%	0%	0%	0.25	0.75
Central Bank	3m EURIBOR (12m_dif	ff) & HP_Diff Lag (t-1yr)		86%	88%	65%	2%	0.10	0.27
Policy	3m EURIBOR (12m_diff	f) & HP_Diff Lag (t-2yr)		57%	75%	16%	0%	0.21	0.63
	Single Indicator: 3	m EURIBOR		64%	80%	33%	0%	0.17	0.50

Table A5: Univariate Benchmark Model - 12m Forecasting Horizon - Out-Of-Sample

Out-Of-Samı	ole			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission Channel	Classification Signal Issued	Recession A C	No Recession B	True positive vs. False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		+ (1 - f) * (Type1Error)
Channel	No Signal Issued 3m EURIBOR	С	D	83%	78%	35%	1%	L= 0.25 0.17	L= 0.75 0.49
	3m EURIBOR (6m_diff) 3m EURIBOR (12m_diff)	f)		43% 67%	69% 68%	4% 0%	0% 0%	0.24	0.72 0.75
Rates	10yr German Governm 10yr German Governm			57% 49%	68% 68%	0%	0% 0%	0.25	0.75 0.75
	10yr German Governm			55%	68%	0%	0%	0.25	0.75
	Term Spread: 10y-3m Term Spread: 10y-3m (	6m_diff)		89% 51%	74% 68%	35% 2%	8% 0%	0.25	0.51 0.74
Term Spread	Term Spread: 10y-3m ( Term Spread: 10y-Swit	12m_diff) ch[3m Euribor/Shadow	Ratel	66% 89%	68% 75%	0% 27%	0% 2%		0.75 0.55
	Term Spread: 10y-Swit	ch[3m Euribor/Shadow   ch[3m Euribor/Shadow	Rate] (6m_diff)	51% 66%	68% 68%	0%	0% 0%	0.22 0.25 0.25 0.25 0.19 0.25 0.13 0.16 0.14 0.09 0.23 0.20 0.23 0.20 0.23 0.20 0.23 0.20 0.25 0.25 0.25 0.25 0.25 0.25 0.25	0.75 0.75
	M1 (6m_ret)	cii[3iii Euriboi/3iiauow	Matej (12111_u111)	77%	84%	55%	3%	0.13	0.35
	M1 (12m_ret) Real M1 (6m_ret)			95% 78%	84% 82%	87% 51%	17% 3%		0.14
	Real M1 (12m_ret) M2 (6m_ret)			93% 83%	89% 72%	78% 20%	5% 3%		0.18 0.61
Money Supply	M2 (12m_ret) Real M2 (6m ret)			91% 81%	76% 71%	35% 18%	4% 3%	0.20	0.50 0.62
	Real M2 (12m_ret)			89%	74%	31%	6%	0.22	0.53
	M3 (6m_ret) M3 (12m_ret)			76% 85%	71% 74%	15% 27%	2% 3%	0.21	0.65 0.55
	Real M3 (6m_ret) Real M3 (12m_ret)			75% 84%	71% 75%	13% 27%	2% 3%		0.66 0.55
	FX EUR/USD			89%	76%	38%	6%	0.20	0.48
	FX EUR/USD (6m_ret) FX EUR/USD (12m_ret)			52% 52%	68% 68%	0% 0%	0% 0%	0.25	0.75 0.75
Competitiveness	JPM Nominal Broad Ef JPM Nominal Broad Ef	fective FX Rate fective FX Rate (6m_ret)		87% 53%	69% 68%	18% 0%	7% 0%		0.63 0.75
competitiveness	JPM Nominal Broad Ef Terms of Trade	fective FX Rate (12m_ret	)	51% 86%	68% 74%	0% 76%	0% 28%	0.25	0.75 0.25
	Terms of Trade (6m_re			50%	68%	0%	0%	0.25	0.75
	Terms of Trade (12m_r PMI	etj		53% 69%	68% 74%	0% 20%	0% 1%	0.21	0.75 0.60
	PMI (6m_diff) PMI (12m_diff)			87% 92%	78% 82%	38% 55%	3% 4%		0.47 0.35
	Economic Sentiment Economic Sentiment (6	5m diff)		50% 91%	68% 82%	0% 51%	0%	0.25	0.75 0.37
	Economic Sentiment (			83%	75%	33%	4%	0.20	0.52
	Industrial Confidence Industrial Confidence	(6m_diff)		44% 92%	68% 79%	0% 47%	0% 5%		0.75 0.41
Sentiment	Industrial Confidence ( Consumer Confidence			83% 54%	71% 68%	20%	4% 0%	0.23	0.61 0.75
	Consumer Confidence	(6m_diff)		80%	78%	44%	5%	0.18	0.44
	Consumer Confidence Consumer Economic Si	ituation		76% 65%	74% 76%	24% 25%	3% 0%	0.19	0.58 0.56
	Consumer Economic Si Consumer Economic Si			82% 80%	76% 79%	40% 42%	6% 3%		0.47 0.44
	Business Climate Business Climate (6m			44% 90%	68% 80%	0% 47%	0% 4%	0.25	0.75 0.41
	Business Climate (12m	_diff)		83%	71%	18%	4%	0.24	0.62
	Corporate Bonds: Inve	stment Grade Spread stment Grade Spread (6:	m_diff)	79% 87%	77% 78%	36% 35%	3% 1%		0.49 0.49
	Corporate Bonds: Inve	stment Grade Spread (12	2m_diff)	94% 70%	84% 72%	53% 16%	2% 1%		0.36 0.63
	High Yield Spread (6m	_diff)		86%	76%	31%	3%	0.19	0.52
Refinancing	High Yield Spread (12n EZ Peripheral Bond Spr	read		95% 76%	84% 74%	55% 53%	2% 16%	0.24	0.35 0.39
	EZ Peripheral Bond Spi EZ Peripheral Bond Spi			81% 84%	74% 82%	29% 55%	4% 4%		0.54 0.35
	MFI Loans to Private Se MFI Loans to Private Se	ector (6m_ret)		66% 76%	68%	0%	0% 3%	0.25	0.75 0.74
	Implied Rates Volatilit	у		64%	75%	24%	1%	0.20	0.57
	Implied Rates Volatility Implied Rates Volatility	y (6m_diff) y (12m_diff)		87% 84%	73% 73%	36% 27%	10% 5%		0.50 0.56
	VSTOXX VSTOXX (6m ret)			64% 71%	68% 68%	2% 7%	1% 3%		0.74 0.70
	VSTOXX (12m_ret)			81%	68%	16%	7%	0.26	0.64
Risk Aversion	VIX VIX (6m_ret)			75% 76%	72% 71%	16% 20%	2% 5%	0.24	0.63 0.61
KISK AVEISION	VIX (12m_ret) FX Basis EUR/USD			92% 82%	78% 76%	44%	6% 6%	0.19	0.44
	FX Basis EUR/USD (6mg FX Basis EUR/USD (12mg			82% 89%	70% 76%	7% 47%	0% 10%	0.23 0.21	0.70 0.42
	Implied FX Vol			54%	71%	11%	0%	0.22	0.67
	Implied FX Vol (6m_ret Implied FX Vol (12m_ret	et)		72% 82%	69% 71%	20% 31%	7% 10%	0.25 0.24	0.62 0.54
	LiborOIS Industrial Production:	Industry excl. Constr. (6r	m ret)	88% 68%	82% 71%	44% 13%	0% 2%	0.14	0.42
		Industry excl. Constr. (12		54% 89%	69% 78%	5% 36%	0%	0.24	0.71 0.48
	Unemployment Rate (	6m_ret)		47%	68%	0%	0%	0.25	0.75
	Unemployment Rate (: Manufact. New Orders	12m_ret) s (6m_ret)		66% 72%	68% 72%	0% 16%	0% 1%	0.25	0.75 0.63
	Manufact. New Orders Industr. Turnover: Mar	(12m_ret)		60% 64%	69% 70%	9% 11%	3% 2%	0.25 0.24	0.69 0.67
Industrial Data	Industr. Turnover: Mar	nufacturing (12m_ret)		53%	68%	0%	0%	0.25	0.75
	MfG Stocks of Finished MfG Stocks of Finished	Products (6m_diff)		54% 93%	70% 77%	7% 47%	0% 9%	0.23 0.20	0.70 0.42
	MfG Stocks of Finished Retail Sales (6m_ret)	Products (12m_diff)		87% 76%	78% 73%	42% 22%	4% 3%	0.18 0.22	0.45 0.59
	Retail Sales (12m_ret) New Passenger Cars (6	m ret)		67%	67%	7%	4%	0.26	0.71
	New Passenger Cars (1			63% 69%	68% 72%	0% 18%	0% 3%	0.25 0.22	0.75 0.62
	HICP (6m_ret) HICP (12m_ret)			64% 61%	68% 69%	0% 5%	0% 0%	0.25 0.24	0.75 0.71
Inflation	HICP excl. Energy, Food	d, Alcohol & Tobacco (6r d, Alcohol & Tobacco (12		64% 68%	68% 68%	0% 0%	0% 0%	0.25 0.25	0.75 0.75
	Domestic PPI: Industry	excluding Construction	(6m_ret)	61%	68%	0%	0%	0.25	0.75
	Crude Oil	excluding Construction	(12m_ret)	66% 81%	68% 78%	2% 71%	0% 18%	0.25 0.21	0.74 0.26
	Crude Oil (6m_diff) Crude Oil (12m_diff)			47% 53%	68% 68%	0%	0% 0%	0.25 0.25	0.75 0.75
Commodities	Gold			81%	56%	53%	42%	0.43	0.46
	Gold (6m_diff) Gold (12m_diff)			49% 57%	68% 68%	0%	0%	0.25	0.75 0.75
	ESTOXX50 (6m_ret) ESTOXX50 (12m_ret)			77% 69%	69% 70%	16% 9%	5% 1%	0.25	0.64
Wealth	Residential Building Pe			93%	75%	35%	5%	0.20	0.50
	Residential Building Pe	ermits (12m_ret)		88%	74%	31%	5%	0.21	0.53

**Table A6: Submodel 1 - Monetary Transmission Channels - Out-Of-Sample** 

Out-Of-Sam	ole			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive rate vs.				L = f * (Type28	Error) + (1 - f) *
Channel	Signal Issued	Α	В	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		
	No Signal Issued	С	D		1		601	L= 0.25	L= 0.75
Channel: Money				94%	89%	80%	<b>6%</b> 5%	0.10	0.17
Real M1 (12r				93%	89%	78%		0.09	0.18
Channel: Compe	itiveness			91%	80%	80%	20%	0.20	0.20
FX EUR/USD				89%	76%	38%	6%	0.20	0.48
Terms of Tra				86%	74%	76%	28%	0.27	0.25
Channel: Sentime				85%	78%	40%	4%	0.18	0.46
PMI (12m_di				92%	82%	55%	4%	0.15	0.35
	ntiment (6m_diff)			91%	82%	51%	3%	0.14	0.37
	nfidence (6m_diff)			92%	79%	47%	5%	0.17	0.41
	nate (6m_diff)			90%	80%	47%	4%	0.16	0.41
	onomic Situation (12m_	diff)		80%	79%	42%	3%	0.17	0.44
	ontidence (6m_ditt)			80%	78%	44%	5%	0.18	0.44
Channel: Refinan				88%	84%	56%	3%	0.14	0.34
	onds: Investment Grade S			79%	77%	36%	3%	0.19	0.49
	Bond Spread (12m_diff	)		84%	82%	55%	4%	0.15	0.35
	read (12m_diff)			95%	84%	55%	2%	0.13	0.35
Channel: Pretere	nce for Risk			80%	76%	31%	2%	0.19	0.52
LiborOIS				88%	82%	44%	0%	0.14	0.42
Implied Rate				64%	75%	24%	1%	0.20	0.57
FX Basis EUR				82%	76%	40%	6%	0.20	0.47
Channel: Industr				86%	85%	62%	3%	0.12	0.30
Retail Sales (				76%	73%	22%	3%	0.22	0.59
	f Finished Products (6m_	_diff)		93%	77%	47%	9%	0.20	0.42
	er Cars (12m_ret)			69%	72%	18%	3%	0.22	0.62
Channel: Inflatio				68%	71%	11%	0%	0.22	0.67
HICP (12m_r				61%	69%	5%	0%	0.24	0.71
Domestic PP	: Industry excluding Con	struction (12m_ret)		66%	68%	2%	0%	0.25	0.74
Channel: Commo	dities			84%	79%	76%	19%	0.20	0.23
Crude Oil				81%	78%	71%	18%	0.21	0.26
Gold				81%	56%	53%	42%	0.43	0.46

**Table A7: Submodel 2 - Bivariate Specification - Out-Of-Sample** 

Out-Of-Samp	ole			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive rate vs.				L = f * (Type2)	Error) + (1 - f) *
Channel	Signal Issued No Signal Issued	A C	B D	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	L= 0.25	L= 0.75
	Real M1 (12m ret) & P	MI (12m diff)		99%	92%	78%	2%	0.07	0.17
	Real M1 (12m_ret) & Ir	vestment Grade Spreads		97%	92%	87%	6%	0.08	0.11
	Real M1 (12m_ret) & To	Real M1 (12m_ret) & Terms of Trade			84%	93%	20%	0.17	0.10
	Real M1 (12m_ret) & Retail Sales (6m_ret)			97%	92%	89%	7%	0.08	0.10
	Real M1 (12m_ret) & C	Real M1 (12m ret) & Crude Oil			85%	91%	17%	0.15	0.11
	PMI (12m_diff) & Inves	tment Grade Spreads		93%	85%	56%	2%	0.13	0.33
	PMI (12m_diff) & Terms of Trade  PMI (12m_diff) & Retail Sales (6m_ret)  PMI (12m_diff) & Crude Oil			100%	82%	96%	24%	0.19	0.09
Single Indicators			93%	83%	53%	3%	0.14	0.36	
		e Oil		99%	86%	95%	17%	0.14	0.08
	Corporate Investment Grade Spreads & Terms of Trade			89%	80%	82%	21%	0.21	0.19
	Corporate Investment	Grade Spreads & Retail S	ales (6m_ret)	79%	80%	36%	2%	0.17	0.48
	Corporate Investment	Grade Spreads & Crude C	)il	93%	81%	89%	22%	0.19	0.14
	Terms of Trade & Retai	l Sales (6m_ret)		90%	81%	82%	19%	0.19	0.18
	Terms of Trade & Crude Oil			86%	77%	76%	23%	0.23	0.23
	Retail Sales (6m_ret) & Crude Oil			87%	80%	75%	18%	0.20	0.24
	Channels: Money Supply & Industrial			97%	94%	91%	4%	0.05	0.08
	Channels: Money Supp	ly & Refinancing		98%	92%	80%	2%	0.07	0.16
	Channels: Money Supp	ly & Competitiveness		97%	87%	91%	15%	0.14	0.11
	Channels: Money Supp	ly & Sentiment		97%	92%	80%	2%	0.07	0.16
	Channels: Money Supp	ly & Commodities		96%	88%	89%	13%	0.12	0.11
	Channels: Industrial &			88%	88%	67%	2%	0.10	0.25
	Channels: Industrial &	Competitiveness		87%	87%	67%	4%	0.11	0.26
Channels	Channels: Industrial &	Sentiment		86%	87%	64%	2%	0.11	0.28
	Channels: Industrial &	Commodities		99%	90%	91%	11%	0.11	0.10
	Channels: Refinancing	& Competitiveness		88%	84%	55%	3%	0.14	0.35
	Channels: Refinancing			86%	85%	55%	2%	0.13	0.34
	Channels: Refinancing			98%	91%	82%	6%	0.09	0.15
	Channels: Competitive			84%	86%	55%	0%	0.11	0.34
	Channels: Competitive	ness & Commodities		95%	77%	84%	25%	0.23	0.19
	Channels: Sentiment &	Commodities		99%	91%	85%	7%	0.09	0.13

Table A8: Submodel 3 - Unstable Surges - Out-Of-Sample

Out-Of-Samp	ole			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive rate vs.				L = f * (Type2Error) +	+ (1 - f) * (Type1Erro
Channel	Signal Issued	A	В	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		
Citatillei	No Signal Issued	C	D		000/		4447	L= 0.25	L= 0.75
	Real M1 & HP_Diff Lag Real M1 & HP_Diff Lag	( (t0)		<b>97%</b> 97%	89% 92%	89% 87%	11% 6%	0.11 0.07	0.11 0.11
Money Supply	Real M1 & HP_Diff Lag			94%	92%	75%	4%	0.07	0.11
	Single Indicator: R			94%	90%	75%	4% 5%	0.09	0.20
	ESTOXX50 (12m ret) &			67%	72%	13%	2%	0.09	0.66
	ESTOXX50 (12III_16t) &			73%	72%	15%	3%	0.23	0.65
Equity Markets				72%	73%	16%	2%	0.24	0.63
	ESTOXX50 (12m_ret) & HP_Diff Lag (t-2yr) ESTOXX50 (12m_ret)			69%	70%	<u> </u>		0.23	0.68
		et) ermits (6m ret) & HP Dif	f Lag (t0)	96%	70%	9% <b>38%</b>	1% 6%	0.23	0.48
		ermits (6m_ret) & HP_Diff		95%	81%	47%	4%	0.20	0.46
<b>Building Permits</b>		ermits (6m_ret) & HP_Diff		92%	78%	36%	4%	0.19	0.41
	Residential Building re	ng Permits (6m ret)	Lag (L-Zyi)	93%	75%	35%	5%	0.19	0.49
	PMI (12m diff) & HP I			100%	86%	56%	1%	0.12	0.33
C	PMI (12m diff) & HP D			92%	83%	53%	4%	0.15	0.36
	PMI (12m_diff) & HP_D			92%	83%	55%	4%	0.13	0.35
	Single Indicator: P			92%	82%	55%	4%	0.14	0.35
	HICP (12m ret) & HP D			64%	51%	20%	36%	0.47	0.69
	HICP (12m_ret) & HP_I			74%	83%	49%	2%	0.15	0.39
Inflation	HICP (12m ret) & HP D			71%	79%	33%	1%	0.17	0.51
	Single Indicator: H			61%	69%	5%	0%	0.24	0.71
	Retail Sales (6m ret) 8			84%	85%	71%	10%	0.14	0.24
	Retail Sales (6m ret) &			78%	78%	36%	3%	0.18	0.49
Retail	Retail Sales (6m ret) &			76%	75%	24%	3%	0.21	0.58
		etail Sales (6m ret)		76%	73%	22%	3%	0.22	0.59
		ector (12m ret) & HP Di	ff Lag (t0)	80%	75%	24%	3%	0.21	0.58
		ector (12m ret) & HP Diff		93%	76%	33%	6%	0.21	0.52
Credit Dynamics		ector (12m_ret) & HP_Diff		82%	78%	29%	1%	0.18	0.53
		AFI Loans to Private Secto		76%	66%	2%	3%	0.27	0.74
	Crude Oil & HP Diff La	g (t0)		89%	76%	78%	25%	0.25	0.23
	Crude Oil & HP Diff Lag	g (t-1yr)		89%	78%	73%	19%	0.21	0.25
Oil	Crude Oil & HP_Diff Lag	g (t-2yr)		91%	80%	78%	20%	0.20	0.21
	Single Indicator: C	rude Oil		81%	78%	71%	18%	0.21	0.26
	3m EURIBOR (12m_diff			74%	70%	0%	0%	0.25	0.75
Central Bank		f) & HP_Diff Lag (t-1yr)		91%	87%	65%	4%	0.12	0.27
Policy	3m EURIBOR (12m_diff	f) & HP_Diff Lag (t-2yr)		66%	70%	0%	0%	0.25	0.75
,	Single Indicator: 3			83%	78%	35%	1%	0.17	0.49

Table A9: The Univariate Benchmark Model over the 6m Forecasting Horizon

None   Company	6m Forecasti	ing Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Part   Color		Signal Issued		В		(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		
Martin   M	C.I.C.III.C.I	3m EURIBOR	•	l D	62%				0.25	0.75
March   The Company   The Co									L=f* (Type2Error)+ L= 0.25 0.25 0.25 0.22 0.25 0.25 0.23 0.25 0.25 0.25 0.27 0.27 0.17 0.10 0.17 0.10 0.17 0.10 0.25 0.25 0.25 0.25 0.25 0.25 0.25 0.2	
Micros   1807	Rates	10yr German Governm	nent Bond				A / (A + C)			
Tem Superson   Tem		10yr German Governm	nent Bond (12m_diff)		62%	77%	0%	0%	L = f* (Type2Error) + (1 - f) * (T   L = 0.25	0.75
Tem Speed   Per   Speed   Company			(6m diff)							
Personal Life Section   1985   775   785   78   78   78   78   78	Term Spread	Term Spread: 10y-3m (	(12m_diff)	2-4-1	60%	77%	0%	0%	0.25	0.75
Miles   Mile	rem spread	Term Spread: 10y-Swit	ch[3m Euribor/Shadow	Rate] Rate] (6m_diff)					L = f* (Type2Error) + (1)	
Miles   1985			ch[3m Euribor/Shadow	Rate] (12m_diff)						
Marcey Supple   Marcey Suppl		M1 (12m_ret)			93%	90%	70%	3%	0.10	0.24
Many Supple   Mail										
Monthly Sept   Rep 100 [Can and 1   255   775   05										
M. J. Gen. red)	Money Supply	Real M2 (6m_ret)			52%	77%	0%	0%	0.25	0.75
Mail   15th										
Red MJ (12m etc.)  Per MJ (12m e		M3 (12m_ret)								0.75
Y. CLUIU LO (Fam. et al. 1975)										
FLUIVISD DIAM red   148   775										
Competitiveness   PM Monimal Book of Riches (m.) reliable   545   775   67		FX EUR/USD (12m_ret)			54%	77%	0%	0%	0.25	0.75
PM Hammal Broad Price of Picke   12m, or   0.5	Competitiveness									
Terms of Trade (Gn ve)		JPM Nominal Broad Ef	fective FX Rate (12m_ret	)					0.25	0.75
PAID   SERVING		Terms of Trade (6m_re	et)		61%	77%	0%	0%	0.25 0.23 0.25 0.18 0.25 0.25 0.25 0.25 0.25 0.27 0.25 0.25 0.25 0.25 0.25 0.25 0.25 0.26 0.27 0.20 0.25 0.21 0.20 0.25 0.21 0.20 0.25 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.27 0.25 0.16 0.10 0.10 0.11 0.13 0.12 0.16 0.18 0.13 0.11 0.11 0.15 0.13 0.12 0.16 0.19 0.10 0.11 0.11 0.15 0.13 0.22 0.11 0.16 0.19 0.16 0.19 0.16 0.19 0.16 0.19 0.11 0.16 0.22 0.19 0.11 0.11 0.14 0.22 0.20 0.25 0.25 0.25 0.25 0.20 0.22 0.20 0.25 0.25	0.75
PAUL Form   1985   85%   97%   35%   0.15   0.44			ret)							
Fecusionis Sentiment   77%   81%   28%   34%   0.20   0.55		PMI (6m_diff)			84%	86%	47%	3%	0.15	0.41
Economic Sentiment [lim, diff]   94%   90%   70%   3%   0.20   0.24		Economic Sentiment			75%	81%	28%	3%	0.20	0.55
Industrial Confidence (Gm Affin										
Sentiment		Industrial Confidence			64%	79%	16%	2%	0.22	0.63
Consume Confidence   838   838   838   828   828   838   8										
Consumer Confidence (12m diff)	Sentiment	Consumer Confidence	<u> </u>		83%	83%	42%	5%	0.18	0.45
Consumer Economic Situation (Em., diff)										
Consumer Economic Stuation (12m) (15m) (										
Buiness Climate (Em. drift)		Consumer Economic S			89%	87%	65%	6%	0.13	0.28
Suines Climate (2m diff)			diff)							
Comporate Bonds: Investment Grade Spread (Em. diff) 75% 85% 35% 0% 0.15 0.48 0.49 0.49 0.49 0.49 0.49 0.49 0.49 0.49		Business Climate (12m	_diff)		89%	84%	47%	5%	0.17	0.41
High Yield Spread (Em. diff)		Corporate Bonds: Inve	stment Grade Spread (6	m_diff)	61%	82%	23%	0%	0.19	0.58
Refinancing High Yield Spread (Em. diff) 80% 90% 58% 1% 0.11 0.32 EX Peripheral Bond Spread (Em. diff) 80% 90% 58% 1% 0.11 0.32 EX Peripheral Bond Spread (Em. diff) 83% 82% 47% 1% 0.01 0.46 EX Peripheral Bond Spread (Em. diff) 83% 82% 47% 1% 0.01 0.46 EX Peripheral Bond Spread (Em. diff) 83% 82% 47% 1% 0.01 0.46 EX Peripheral Bond Spread (Em. diff) 83% 82% 100% 3% 0.02 0.33 MH Loans to Private Sector (Idm ret) 54% 77% 0% 0% 0.45 0.75 0.75 MH Loans to Private Sector (Idm ret) 54% 77% 0% 0% 0.42 0.50 0.75 0.75 Implied Retex Volatility (Em. diff) 83% 83% 81% 35% 6% 0.00 0.50 0.50 0.50 0.50 0.50 0.50 0.5			stment Grade Spread (1	2m_diff)						
E Peripheral Bond Spread [ 77%   87%   47%   19%   0.14   0.40		High Yield Spread (6m)	_diff)		67%	82%	28%	1%	0.19	0.54
## EZ Peripheral Bond Spread (1zm_diff)	Refinancing	EZ Peripheral Bond Sp	n_diff) read							
MFI Loans to Private Sector (Em. ret)		EZ Peripheral Bond Sp	read (6m_diff)							
Implied Rates Volatility (in: diff)		MFI Loans to Private S	ector (6m_ret)		54%	77%	0%	0%	0.25	0.75
Implied Rates Volatility (Em., diff)										
VSTOXX		Implied Rates Volatilit	y (6m_diff)		63%	80%	12%	0%	0.22	0.66
NSTOXX (12m, ret)		VSTOXX	y (12m_airi)							
VIX										
VIX (12m ret)		VIX			77%	79%	23%	5%	0.23	0.59
F R Basis EUR/USD (em diff)	Risk Aversion									
FX Basis EUR/USO (12m diff)		FX Basis EUR/USD	4160				40%		0.19	0.47
Implied FX Vol (6m_ret)										
Implied FX Vol (12m ret)		Implied FX Vol (6m re	t)							
Industrial Production: Industry excl. Constr. (6m_ret)		Implied FX Vol (12m_r	et)		68%	78%	14%	3%	0.24	0.65
Unemployment Rate   54%   77%   0%   0%   0.25   0.75		Industrial Production:								
Unemployment Rate (cm, ret)		Industrial Production:				77%	12%		0.25	0.67
Manufact. New Orders (6m_ret)   80%   84%   35%   1%   0.17   0.49		Unemployment Rate (	6m_ret)		79%	79%	16%	3%	0.23	0.63
Industrial Data   Industrial		Unemployment Rate ( Manufact, New Orders	12m_ret) s (6m_ret)							
Industrial Data		Manufact. New Orders	s (12m_ret)		70%	78%	14%	3%	0.24	0.65
MfG Stocks of Finished Products (and diff)         63%         80%         14%         0%         0.22         0.65           MfG Stocks of Finished Products (and diff)         85%         86%         47%         3%         0.15         0.36           Retail Sales (and products (and fill)         89%         86%         53%         5%         0.15         0.36           Retail Sales (and pret)         90%         86%         51%         4%         0.15         0.36           Retail Sales (12m ret)         87%         82%         47%         8%         0.19         0.42           New Passenger Cars (6m ret)         75%         75%         78%         12%         3%         0.24         0.67           New Passenger Cars (12m ret)         80%         81%         33%         5%         0.21         0.52           HICP (6m ret)         72%         79%         7%         0%         0.23         0.70           HICP (6m ret)         72%         79%         7%         0%         0.23         0.70           HICP (6m ret)         72%         79%         7%         0%         0.23         0.70           HICP (6m ret)         72%         79%         7%         0%	Industrial Data	Industr. Turnover: Mai	nufacturing (12m_ret)		52%	77%	0%	0%	0.25	0.75
MfG Stocks of Finished Products (12m_diff)   89%   86%   53%   5%   0.15   0.36		MfG Stocks of Finished	l Products		63%	80%	14%		0.22	0.65
Retail Sales (12m_ret)   87%   82%   47%   8%   0.19   0.42		MfG Stocks of Finished			89%	86%	53%	5%	0.15	0.36
New Passenger Cars (6m_ret)   75%   78%   12%   3%   0.24   0.67		Retail Sales (6m_ret) Retail Sales (12m_ret)				86% 82%				
HiCP (6m, ret)   72%   79%   7%   0%   0.23   0.70		New Passenger Cars (6			75%	78%	12%	3%	0.24	0.67
HiCP excl. Energy, Food, Alcohol & Tobacco (6m_ret)   69%   74%   0%   4%   0.28   0.76     HiCP excl. Energy, Food, Alcohol & Tobacco (12m_ret)   70%   72%   0%   6%   0.30   0.77     Domestic PPI: Industry excluding Construction (6m_ret)   68%   77%   0%   0%   0.25   0.75     Domestic PPI: Industry excluding Construction (12m_ret)   78%   81%   28%   3%   0.20   0.55     Crude Oil   28%   80%   37%   8%   0.21   0.49     Crude Oil (6m_diff)   53%   77%   0%   0%   0.25   0.75     Crude Oil (6m_diff)   51%   77%   0%   0%   0.25   0.75     Gold (12m_diff)   56%   77%   0%   0%   0.25   0.75     Gold (12m_diff)   56%   77%   0%   0%   0.25   0.75     Gold (12m_diff)   56%   77%   0%   0%   0.25   0.75     ESTOXISO (6m_ret)   74%   79%   23%   5%   0.23   0.59     ESTOXISO (12m_ret)   77%   78%   21%   6%   0.24   0.61     Residential Building Permits (6m_ret)   78%   84%   40%   3%   0.17   0.46		HICP (6m_ret)	zm_let/		72%	79%	7%	0%	0.23	0.70
HiCP ext. Energy, Food, Alcohol & Tobacco (12m ret)   70%   72%   0%   6%   0.30   0.77			d. Alcohol & Tohacco (6r	n ret)						
Domestic PPI: Industry excluding Construction (12m_ret)	Inflation	HICP excl. Energy, Food	d, Alcohol & Tobacco (12	m_ret)	70%	72%	0%	6%	0.30	0.77
Commodities         Crude Oil (6m, diff)         53%         77%         0%         0%         0.21         0.49           Commodities         Crude Oil (12m diff)         53%         77%         0%         0%         0.25         0.75           Gold         76%         88%         51%         1%         0.13         0.37           Gold (6m_diff)         56%         77%         0%         0%         0.25         0.75           BSTOXX50 (6m_ret)         56%         77%         0%         0%         0.25         0.75           Wealth         ESTOXX50 (6m_ret)         74%         79%         23%         5%         0.23         0.59           Residential Building Permits (6m_ret)         77%         78%         21%         6%         0.24         0.61           84         40%         3%         0.17         0.46										
Commodities         Crude Oil (12m_diff)         51%         77%         0%         0%         0.25         0.75           Gold Gold (Gold (		Crude Oil	0		82%	80%	37%	8%	0.21	0.49
Gold (6m_diff)   56%   77%   0%   0%   0.25   0.75	Commodition	Crude Oil (12m_diff)			51%	77%	0%	0%	0.25	0.75
Gold (12m diff)         60%         77%         0%         0%         0.25         0.75           ESTOXX50 (6m_ret)         74%         79%         23%         5%         0.23         0.59           ESTOXX50 (12m_ret)         77%         78%         21%         6%         0.24         0.61           Residential Building Permits (6m_ret)         78%         84%         40%         3%         0.17         0.46	commodities	Gold							0.13	0.37
Wealth         ESTOXX50 (12m_ret)         77%         78%         21%         6%         0.24         0.61           Residential Building Permits (6m_ret)         78%         84%         40%         3%         0.17         0.46		Gold (12m_diff)			60%	77%	0%	0%	0.25	0.75
Wealth         Residential Building Permits (6m_ret)         78%         84%         40%         3%         0.17         0.46										
	Wealth	Residential Building Pe								

Table A10: The Univariate Benchmark Model over the 18m Forecasting Horizon

18m Forecas	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm		Risk-Averse
Transmission Channel	Classification Signal Issued No Signal Issued	Recession A C	No Recession B D	True positive vs. False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	L = f * (Type2Error) - L= 0.25	+ (1 - f) * (Type1Erro
	3m EURIBOR		В	65%	69%	39%	13%	0.25	0.49
D-4	3m EURIBOR (6m_d 3m EURIBOR (12m_	diff)		38% 68%	62% 68%	0% 22%	0% 4%	0.25 0.22	0.75 0.59
Rates	10yr German Gover	nment Bond nment Bond (6m_diff)		52% 54%	62% 63%	0% 3%	0% 0%	0.25 0.24	0.75 0.73
	10yr German Gover	nment Bond (12m_diff)		51%	62%	0%	0%	0.25	0.75
	Term Spread: 10y-3 Term Spread: 10y-3	m m (6m diff)		70% 56%	70% 62%	46% 0%	15% 0%	0.25 0.25	0.44 0.75
Term Spread	Term Spread: 10y-3	m (12m_diff)	intal	71%	66%	25%	9%	0.26	0.58
·	Term Spread: 10y-St	witch[3m Euribor/Shadow F witch[3m Euribor/Shadow F	tate] (6m_diff)	84% 66%	76% 69%	61% 18%	15% 0%	0.21 0.21	0.33 0.62
	Term Spread: 10y-St M1 (6m_ret)	witch[3m Euribor/Shadow F	tate] (12m_diff)	79% 84%	76% 79%	67% 66%	19% 13%	0.22	0.29
	M1 (12m_ret)			89%	79%	73%	17%	0.19	0.24
	Real M1 (6m_ret) Real M1 (12m_ret)			84% 90%	78% 81%	67% 75%	15% 15%	0.19 0.17	0.28 0.23
	M2 (6m_ret) M2 (12m_ret)			52% 52%	62% 62%	0% 0%	0% 0%	0.25 0.25	0.75 0.75
Money Supply	Real M2 (6m_ret)			50%	62%	0%	0%	0.25	0.75
	Real M2 (12m_ret) M3 (6m_ret)			49% 50%	62% 62%	0% 0%	0% 0%	0.25 0.25	0.75 0.75
	M3 (12m_ret)			49%	62%	0%	0%	0.25	0.75
	Real M3 (6m_ret) Real M3 (12m_ret)			52% 51%	62% 62%	0% 0%	0% 0%	0.25 0.25	0.75 0.75
	FX EUR/USD FX EUR/USD (6m_re	+1		82% 54%	70% 64%	54% 6%	20% 0%	0.27 0.24	0.40 0.71
	FX EUR/USD (12m_r			54%	62%	0%	0%	0.25	0.75
`amnatitiyanass	JPM Nominal Broad	Effective FX Rate Effective FX Rate (6m_ret)		64% 61%	63% 59%	24% 10%	12% 10%	0.28 0.30	0.60 0.70
Lompetitiveness	JPM Nominal Broad	Effective FX Rate (12m_ret		64%	63%	28%	16%	0.30	0.58
	Terms of Trade Terms of Trade (6m)	ret)		82% 64%	74% 64%	66% 12%	20% 4%	0.24 0.25	0.31 0.67
	Terms of Trade (12n	n_ret)		72%	65%	49%	25%	0.31	0.44
	PMI (6m_diff)			57% 76%	71% 73%	30% 51%	4% 14%	0.20	0.54
	PMI (12m_diff)			78%	74%	51% 51% 0%	12%	0.21	0.40
	Economic Sentimen	t (6m_diff)		48% 73%	62% 75%	51%	9%	0.25 0.19	0.75 0.39
	Economic Sentimen			64% 60%	72% 62%	36% 0%	6% 0%	0.20 0.25	0.50
	Industrial Confidence	ce (6m_diff)		70%	74%	45%	7%	0.19	0.75 0.43
Sentiment	Industrial Confident Consumer Confiden			61% 62%	64% 69%	12% 28%	4% 6%	0.25 0.23	0.67 0.55
	Consumer Confiden	ce (6m_diff)		71%	71%	45%	12%	0.23	0.44
	Consumer Confident Consumer Economic			62% 67%	70% 72%	42% 49%	12% 14%	0.24	0.47
	Consumer Economi	c Situation (6m_diff)		73%	70%	48%	17%	0.26	0.43
	Business Climate	c Situation (12m_diff)		69% 60%	70% 62%	45% 0%	14%	0.24	0.45 0.75
	Business Climate (6)	m_diff)		71%	74%	45% 10%	8%	0.20	0.44
	Business Climate (1: Corporate Bonds: In	vestment Grade Spread		62% 82%	64% 83%	69%	3% 8%	0.24 0.14	0.68 0.26
		vestment Grade Spread (6r vestment Grade Spread (12		64% 72%	72% 77%	30% 43%	2% 3%	0.19 0.16	0.53 0.43
	High Yield Spread		in_uii)	69%	66%	33%	13%	0.27	0.54
Refinancing	High Yield Spread (6 High Yield Spread (1			64% 68%	71% 76%	33% 45%	5% 5%	0.20 0.17	0.52 0.43
nemeneng	EZ Peripheral Bond	Spread		73%	73%	45%	10%	0.21	0.44
	EZ Peripheral Bond EZ Peripheral Bond			80% 88%	78% 81%	45% 55%	1% 3%	0.15 0.13	0.42
	MFI Loans to Private	e Sector (6m_ret)		54%	62%	0%	0%	0.25	0.75
	MFI Loans to Private Implied Rates Volat	ility		60% 74%	68% 74%	30% 51%	8% 11%	0.24	0.55 0.40
	Implied Rates Volat Implied Rates Volat	ility (6m_diff)		63% 62%	69% 66%	33% 22%	8% 6%	0.23 0.24	0.52 0.60
	VSTOXX	mity (12m_dm)		63%	62%	12%	7%	0.28	0.68
	VSTOXX (6m_ret) VSTOXX (12m_ret)			57% 64%	61% 61%	7% 15%	6% 10%	0.27 0.29	0.71 0.66
	VIX			69%	64%	28%	14%	0.28	0.57
Risk Aversion	VIX (6m_ret) VIX (12m_ret)			59% 68%	65% 69%	24% 37%	9% 11%	0.26 0.24	0.59 0.50
	FX Basis EUR/USD	im diff)		77%	71%	55%	19%	0.25	0.38
	FX Basis EUR/USD (1			70%	74%	37%	4%	0.18	0.48
	Implied FX Vol Implied FX Vol (6m_	ret)		62% 56%	70% 65%	30% 15%	5% 4%	0.21	0.54 0.65
	Implied FX Vol (12m	_ret)		59%	65%	16%	6%	0.25	0.64
		n: Industry excl. Constr. (6n		75% 53%	79% 65%	55% 10%	6% 2%	0.16 0.24	0.35 0.68
		n: Industry excl. Constr. (12		60% 61%	62% 74%	0% 31%	0%	0.25 0.17	0.75 0.51
	Unemployment Rat	e (6m_ret)		58%	64%	10%	3%	0.24	0.68
	Unemployment Rat Manufact. New Ord			51% 53%	62% 65%	0% 10%	0% 1%	0.25 0.23	0.75 0.67
	Manufact. New Ord	ers (12m_ret)		57%	62%	0%	0%	0.25	0.75
Industrial Data		Manufacturing (6m_ret) Manufacturing (12m_ret)		43% 70%	62% 68%	0% 19%	0% 2%	0.25 0.22	0.75 0.61
	MfG Stocks of Finish	ned Products		64%	66%	10%	0%	0.22	0.67
	MfG Stocks of Finish	ned Products (6m_diff) ned Products (12m_diff)		69% 63%	71% 69%	43% 22%	11% 3%	0.23 0.21	0.45 0.59
	Retail Sales (6m_ret Retail Sales (12m re	)		75% 70%	71% 71%	48% 45%	14% 13%	0.23 0.24	0.43 0.45
	New Passenger Cars	(6m_ret)		73%	71%	42%	11%	0.23	0.46
	New Passenger Cars HICP (6m ret)	(12m_ret)		80% 70%	75% 71%	54% 37%	12% 7%	0.21 0.21	0.38 0.49
	HICP (12m_ret)		.)	75%	67%	52%	23%	0.29	0.42
Inflation		ood, Alcohol & Tobacco (6n ood, Alcohol & Tobacco (12		66% 63%	59% 54%	15% 0%	14% 12%	0.32 0.34	0.67 0.78
	Domestic PPI: Indus	try excluding Construction	6m_ret)	74%	66%	31%	12%	0.26	0.55
	Domestic PPI: Indus Crude Oil	try excluding Construction	12m_ret)	81% 79%	70% 74%	57% 61%	22% 19%	0.27	0.38
	Crude Oil (6m_diff)	1		53%	62%	0%	0%	0.25	0.75
Commodities	Crude Oil (12m_diff Gold	)		50% 76%	62% 65%	0% 49%	0% 25%	0.25	0.75 0.44
	Gold (6m_diff)			58%	62%	7%	4%	0.26	0.70
	Gold (12m_diff) ESTOXX50 (6m_ret)			67% 64%	65% 66%	28% 27%	12% 10%	0.27	0.57 0.57
Wealth	ESTOXX50 (12m_ret			65%	60%	24%	18%	0.32	0.61
	Residential Building	Permits (6m_ret) Permits (12m_ret)		69% 68%	67% 65%	40% 30%	16%	0.27 0.27	0.49 0.56

**Table A11: The Univariate Benchmark Model over the 24m Forecasting Horizon** 

4m Forecas	sting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm		Risk-Avers
Transmission Channel	Classification Signal Issued	Recession A	No Recession B	True positive vs. False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	L = f * (Type2Error)	
Chamici	No Signal Issued 3m EURIBOR	С	D	63%	59%	42%	27%	L= 0.25 0.35	L= 0.75 0.50
	3m EURIBOR (6m_diff) 3m EURIBOR (12m_diff)			64% 68%	53% 75%	0% 59%	0% 12%	0.25 0.19	0.75 0.33
Rates	10yr German Governme	ent Bond		51%	48%	19%	27%	0.40	0.67
	10yr German Governme 10yr German Governme			50% 57%	56% 56%	5% 37%	0% 28%	0.24	0.71 0.54
	Term Spread: 10y-3m			67%	64%	53%	26%	0.31	0.42
	Term Spread: 10y-3m (6 Term Spread: 10y-3m (1			55% 64%	60% 63%	15% 56%	0% 30%	0.21	0.64
Term Spread	Term Spread: 10y-Switc	h[3m Euribor/Shadow F		80%	75%	70%	21%	0.23	0.28
	Term Spread: 10y-Switc Term Spread: 10y-Switc	n[3m Euribor/Shadow F h[3m Euribor/Shadow F	rate] (6m_diff)	64% 73%	62% 71%	43% 71%	22% 29%	0.31	0.48
	M1 (6m_ret)	•		79%	70%	67%	27%	0.28	0.31
	M1 (12m_ret) Real M1 (6m_ret)			80% 79%	74% 70%	77% 67%	29% 28%	0.27	0.24
	Real M1 (12m_ret)			81%	73%	77%	30%	0.28	0.25
	M2 (6m_ret) M2 (12m_ret)			50% 50%	53% 53%	0% 0%	0%	0.25 0.25	0.75 0.75
Money Supply	Real M2 (6m_ret)			53% 48%	53% 53%	0% 0%	0% 0%	0.25 0.25	0.75 0.75
	Real M2 (12m_ret) M3 (6m_ret)			52%	59%	16%	3%	0.23	0.63
	M3 (12m_ret) Real M3 (6m_ret)			51% 53%	63% 60%	25% 29%	3% 13%	0.21 0.28	0.57 0.56
	Real M3 (12m_ret)			52%	62%	29%	9%	0.24	0.55
	FX EUR/USD FX EUR/USD (6m_ret)			84% 51%	72% 54%	70% 13%	27% 10%	0.28 0.29	0.29 0.68
	FX EUR/USD (12m_ret)			51%	53%	0%	0%	0.25	0.75
	JPM Nominal Broad Effe	ective FX Rate		69%	63%	57%	32%	0.35	0.40
ornpetitiveness	JPM Nominal Broad Effe JPM Nominal Broad Effe	ective FX Rate (bm_ret)		57% 62%	53% 54%	32% 41%	29% 33%	0.39	0.58 0.53
	Terms of Trade Terms of Trade (6m ret	· -		79% 66%	70% 61%	68% 54%	28%	0.29 0.36	0.31
	Terms of Trade (12m_re			71%	68%	65%	29%	0.31	0.34
	PMI			48%	62%	30%	10%	0.25	0.55
	PMI (6m_diff) PMI (12m_diff)			68% 66%	64% 66%	56% 52%	28% 22%	0.32	0.40
	Economic Sentiment	m diff)		45%	53%	0%	0%	0.25	0.75
	Economic Sentiment (6) Economic Sentiment (1)			62% 53%	64% 65%	49% 35%	23% 9%	0.30 0.23	0.44
	Industrial Confidence			60%	73%	43%	1%	0.15	0.43
	Industrial Confidence (6 Industrial Confidence (1			59% 51%	68% 56%	46% 6%	12% 0%	0.23 0.23	0.44
Sentiment	Consumer Confidence	_ ·		59%	63%	48%	23%	0.30	0.45
	Consumer Confidence ( Consumer Confidence (			63% 53%	61% 62%	41% 43%	21% 22%	0.31	0.50
	Consumer Economic Sit			61%	64%	48%	22%	0.30	0.44
	Consumer Economic Sit Consumer Economic Sit	uation (6m_diff) uation (12m_diff)		66%	62% 61%	47% 43%	26% 23%	0.32 0.32	0.46
	Business Climate			61%	72%	41%	1%	0.16	0.45
	Business Climate (6m_d Business Climate (12m_			59% 53%	67% 57%	46% 8%	14% 0%	0.24	0.44
	Corporate Bonds: Invest	tment Grade Spread		79%	82%	73%	10%	0.14	0.22
	Corporate Bonds: Invest Corporate Bonds: Invest	tment Grade Spread (6r tment Grade Spread (12	n_diff) !m_diff)	60% 66%	65% 68%	41% 42%	13% 9%	0.25 0.21	0.48
	High Yield Spread			65%	63%	46%	21%	0.29	0.46
Refinancing	High Yield Spread (6m_ High Yield Spread (12m	diff) diff)		59% 61%	59% 69%	34% 42%	19% 7%	0.31	0.54
	EZ Peripheral Bond Spre	ad		70%	60%	43%	24%	0.33	0.49
	EZ Peripheral Bond Spre EZ Peripheral Bond Spre	ead (6m_diff)		81% 83%	78% 79%	57% 59%	3% 4%	0.13 0.13	0.33 0.31
	MFI Loans to Private See	ctor (6m_ret)		56%	59%	39%	23%	0.33	0.51
	MFI Loans to Private Ser Implied Rates Volatility			61% 71%	57% 68%	43% 52%	30% 18%	0.37	0.50 0.41
	Implied Rates Volatility	(6m_diff)		57%	59%	37%	22%	0.32	0.53
	Implied Rates Volatility VSTOXX	(12m_diff)		55% 62%	53% 56%	29%	26% 19%	0.37	0.60
	VSTOXX (6m_ret)			57%	57%	27%	16%	0.30	0.59
	VSTOXX (12m_ret)			60%	59% 63%	38% 46%	22%	0.32	0.52
Risk Aversion	VIX (6m_ret)			57%	63%	35%	13%	0.26	0.52
AVE: 51011	VIX (12m_ret) FX Basis EUR/USD			62% 77%	63% 72%	43% 66%	20% 22%	0.29	0.48 0.31
	FX Basis EUR/USD (6m_c			56%	63%	22%	1%	0.20	0.59
	FX Basis EUR/USD (12m] Implied FX Vol	_diff)		66% 60%	63% 63%	37% 44%	13% 20%	0.26	0.51 0.47
	Implied FX Vol (6m_ret)			53%	57%	20%	11%	0.28	0.63
	Implied FX Vol (12m_ret LiborOIS	t)		50% 65%	56% 73%	16% 49%	9% 6%	0.28 0.17	0.65 0.39
	Industrial Production: Ir			43%	57%	10%	2%	0.24	0.68
	Industrial Production: In Unemployment Rate	ndustry excl. Constr. (12	m_ret)	62% 62%	62% 44%	19% 46%	0% 58%	0.20 0.57	0.61 0.55
	Unemployment Rate (6)			55%	67%	38%	8%	0.21	0.48
	Unemployment Rate (1: Manufact. New Orders			51% 45%	57% 61%	13% 18%	4% 1%	0.25 0.21	0.67 0.62
	Manufact. New Orders	(12m_ret)		61%	70%	42%	4%	0.18	0.45
ndustrial Data	Industr. Turnover: Mani Industr. Turnover: Mani			65% 70%	54% 70%	3% 59%	0% 20%	0.24 0.25	0.73 0.35
	MfG Stocks of Finished	Products		69%	62%	51%	29%	0.34	0.44
	MfG Stocks of Finished I MfG Stocks of Finished I			58% 49%	62% 54%	38% 3%	18% 0%	0.29 0.24	0.51 0.73
	Retail Sales (6m_ret)	Judeto (±zIII_UIII)		71%	69%	58%	22%	0.27	0.37
	Retail Sales (12m_ret) New Passenger Cars (6n	n ret)		69% 73%	66% 69%	54% 59%	23% 22%	0.29	0.40
	New Passenger Cars (12	m_ret)		72%	69%	63%	27%	0.29	0.34
	HICP (6m_ret) HICP (12m_ret)			68% 71%	62% 56%	61% 58%	38%	0.38	0.39
IEl	HICP excl. Energy, Food,	, Alcohol & Tobacco (6n	n_ret)	71% 59%	56% 50%	58% 28%	46% 30%	0.45	0.43
Inflation	HICP excl. Energy, Food,	, Alcohol & Tobacco (12	m_ret)	55%	46%	0%	14%	0.36	0.79
	Domestic PPI: Industry e Domestic PPI: Industry e			75% 79%	69% 73%	66% 73%	29% 27%	0.30	0.33 0.27
	Crude Oil	o constituction	=0	79%	75%	71%	22%	0.24	0.27
	Crude Oil (6m_diff) Crude Oil (12m_diff)			50% 55%	54% 52%	3% 30%	0% 29%	0.24	0.73 0.59
Commodities	Gold			75%	57%	52%	39%	0.41	0.46
	Gold (6m_diff) Gold (12m_diff)			68% 77%	64% 72%	54% 67%	28% 24%	0.32 0.27	0.41 0.31
	ESTOXX50 (6m_ret)			62%	72% 58%	37%	23%	0.27	0.31
	ESTOXX50 (12m_ret) Residential Building Per	mite (Cm+1		58%	57%	34%	23%	0.34	0.55
Wealth				67%	62%	48%	27%	0.33	0.46

**Table A12: Submodel 1 - Monetary Transmission Channels over Various Time Horizons** 

# 6 months

6m Forecast	ing Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
om rorcease	. —			AOROC	Correct Signar	CH313 Detected	Taise Alaitii		
Transmission	Classification	Recession	No Recession	True positive vs.				L = f * (Type2E	rror) + (1 - f) *
Channel	Signal Issued No Signal Issued	A C	B D	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	L= 0.25	L= 0.75
Channel: Money				95%	92%	72%	2%	0.09	0.21
Real M1 (12r				94%	91%	70%	2%	0.09	0.23
Channel: Compet	itiveness			92%	86%	65%	8%	0.15	0.28
FX EUR/USD				81%	81%	40%	6%	0.20	0.47
Terms of Tra				88%	83%	63%	10%	0.17	0.31
Channel: Sentime				93%	90%	70%	4%	0.11	0.24
PMI (12m_di				95%	90%	74%	5%	0.10	0.20
	ntiment (6m_diff)			94%	90%	70%	3%	0.10	0.24
	nfidence (6m_diff)			92%	89%	65%	4%	0.12	0.27
	nate (6m_diff)			93%	89%	65%	4%	0.12	0.27
	onomic Situation (12m_	diff)		89%	87%	65%	6%	0.13	0.28
	onfidence (6m_diff)			89%	88%	63%	5%	0.13	0.29
Channel: Refinan				89%	88%	56%	3%	0.13	0.34
	onds: Investment Grade			91%	85%	51%	5%	0.16	0.38
	Bond Spread (12m_diff	)		83%	82%	30%	3%	0.20	0.53
	read (12m_diff)			80%	90%	58%	1%	0.11	0.32
Channel: Prefere	nce for Risk			93%	88%	60%	4%	0.13	0.31
LiborOIS				90%	87%	51%	2%	0.14	0.37
Implied Rate				83%	81%	35%	6%	0.20	0.50
FX Basis EUR				80%	82%	40%	6%	0.19	0.47
Channel: Industr				95%	90%	72%	4%	0.10	0.22
Retail Sales (				90%	86%	51%	4%	0.15	0.38
	f Finished Products (6m	_diff)		85%	86%	47%	3%	0.15	0.41
	er Cars (12m_ret)			80%	81%	33%	5%	0.21	0.52
Channel: Inflatio				86%	87%	49%	2%	0.14	0.39
HICP (12m_r				86%	86%	37%	0%	0.16	0.47
	: Industry excluding Con	struction (12m_ret)		78%	81%	28%	3%	0.20	0.55
Channel: Commo	dities			83%	85%	42%	2%	0.16	0.44
Crude Oil				82%	80%	37%	8%	0.21	0.49
Gold				76%	88%	51%	1%	0.13	0.37

# 18 months

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18m Forecas	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive vs.				L = f * (Type28	Frror) + (1 - f) *
Channel	Signal Issued	А	В	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		
	No Signal Issued	С	D					L= 0.25	L= 0.75
Channel: Money				90%	81%	75%	15%	0.17	0.23
Real M1 (12)				90%	81%	75%	15%	0.17	0.23
Channel: Compe	itiveness			85%	74%	66%	20%	0.24	0.31
FX EUR/USD				82%	70%	54%	20%	0.27	0.40
Terms of Tra				82%	74%	66%	20%	0.24	0.31
Channel: Sentime				69%	76%	48%	6%	0.18	0.41
PMI (12m_di				78%	74%	51%	12%	0.21	0.40
	ntiment (6m_diff)			73%	75%	51%	9%	0.19	0.39
	nfidence (6m_diff)			70%	74%	45%	7%	0.19	0.43
	nate (6m_diff)			71%	74%	45%	8%	0.20	0.44
Consumer Ed	onomic Situation (12m_	_diff)		69%	70%	45%	14%	0.24	0.45
Consumer Co	intidence (6m_ditt)			71%	71%	45%	12%	0.23	0.44
Channel: Refinar				83%	83%	66%	6%	0.13	0.27
	onds: Investment Grade			82%	83%	69%	8%	0.14	0.26
	I Bond Spread (12m_diff	ř)		88%	81%	55%	3%	0.13	0.34
High Yield Sp	read (12m_diff)			68%	76%	45%	5%	0.17	0.43
Channel: Prefere	nce for Risk			79%	81%	61%	6%	0.15	0.31
LiborOIS				75%	79%	55%	6%	0.16	0.35
Implied Rate	s Volatility			74%	74%	51%	11%	0.21	0.40
FX Basis EUR				77%	71%	55%	19%	0.25	0.38
Channel: Industr	al Data			82%	79%	60%	9%	0.17	0.33
Retail Sales (	6m_ret)			75%	71%	48%	14%	0.23	0.43
MtG Stocks o	f Finished Products (6m)	_diff)		69%	71%	43%	11%	0.23	0.45
New Passeng	er Cars (12m_ret)			80%	75%	54%	12%	0.21	0.38
Channel: Inflatio				81%	71%	58%	21%	0.26	0.37
HICP (12m_r				75%	67%	52%	23%	0.29	0.42
	: Industry excluding Con	struction (12m_ret)		81%	70%	57%	22%	0.27	0.38
Channel: Commo	dities			83%	74%	70%	23%	0.25	0.28
Crude Oil				79%	74%	61%	19%	0.24	0.34
Gold				76%	65%	49%	25%	0.31	0.44

#### 24 months

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24m Forecas	sting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive vs.				L = f * (Type28	rror) + (1 - f) *
Channel	Signal Issued No Sianal Issued	A C	B D	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	1=0.25	1=0.75
Channel: Money			<u> </u>	79%	72%	76%	31%	0.29	0.26
Real M1 (12r				81%	73%	77%	30%	0.28	0.25
Channel: Compet	itiveness			84%	73%	73%	28%	0.27	0.27
FX EUR/USD				84%	72%	70%	27%	0.28	0.29
Terms of Tra	de			79%	70%	68%	28%	0.29	0.31
Channel: Sentime	ent			58%	64%	47%	20%	0.28	0.45
PMI (12m_di	ff)			66%	66%	52%	22%	0.29	0.42
Economic Se	ntiment (6m_diff)			62%	64%	49%	23%	0.30	0.44
Industrial Co	ntidence (6m_ditt)			59%	68%	46%	12%	0.23	0.44
Business Clin	nate (6m_diff)			59%	67%	46%	14%	0.24	0.44
Consumer Ec	onomic Situation (12m	diff)		60%	61%	43%	23%	0.32	0.49
Consumer Co	onfidence (6m_diff)			63%	61%	41%	21%	0.31	0.50
Channel: Refinan	cing			76%	76%	59%	10%	0.18	0.33
Corporate Bo	onds: Investment Grade	Spread		79%	82%	73%	10%	0.14	0.22
EZ Periphera	l Bond Spread (12m_diff	f)		83%	79%	59%	4%	0.13	0.31
	read (12m_diff)			61%	69%	42%	7%	0.20	0.45
Channel: Prefere	nce for Risk			75%	77%	65%	12%	0.18	0.30
LiborOIS				65%	73%	49%	6%	0.17	0.39
Implied Rate	s Volatility			71%	68%	52%	18%	0.25	0.41
FX Basis EUR,				77%	72%	66%	22%	0.25	0.31
Channel: Industri	al Data			69%	72%	58%	17%	0.23	0.35
Retail Sales (				71%	69%	58%	22%	0.27	0.37
MfG Stocks o	f Finished Products (6m)	_diff)		58%	62%	38%	18%	0.29	0.51
	ger Cars (12m_ret)			72%	69%	63%	27%	0.29	0.34
Channel: Inflation	n			78%	73%	73%	28%	0.27	0.27
HICP (12m_re				71%	56%	58%	46%	0.45	0.43
	: Industry excluding Con	struction (12m_ret)		79%	73%	73%	27%	0.27	0.27
Channel: Commo	dities			83%	79%	82%	23%	0.22	0.19
Crude Oil				79%	75%	71%	22%	0.24	0.27
Gold				75%	57%	52%	39%	0.41	0.46

**Table A13: Submodel 2 - Bivariate Specification over Various Time Horizons** 

# 6 months

6m Forecast	ing Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive vs.				L = f * (Type2E	irror) + (1 - f) *
Channel	Signal Issued	A	В	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		
Chamici	No Signal Issued	C	D		050/	240/	20/	L= 0.25	L= 0.75
	Real M1 (12m_ret) & P			99% 98%	95% 94%	84% 84%	2% 3%	0.06	0.13 0.13
	Real M1 (12m_ret) & Investment Grade Spreads Real M1 (12m_ret) & Terms of Trade			95%		72%	3%	0.07	0.13
				95%	91%	81%	3%	0.09	0.22
	Real M1 (12m_ret) & R Real M1 (12m_ret) & C			99%	94% 91%	81% 70%	3%	0.07	0.15
				98%	91%	70%	2%	0.10	0.23
	PMI (12m_diff) & Inves			100%	98%	93%	1%	0.07	0.18
	PMI (12m_diff) & Term			97%	98%	81%	6%		0.05
Single Indicators	PMI (12m_diff) & Retai							0.09	
	PMI (12m_diff) & Crude		·-	100%	97%	93%	1%	0.03	0.06
		Grade Spreads & Terms of		95%	91%	79%	5%	0.09	0.17
	Corporate Investment Grade Spreads & Retail Sales (6m_ret) Corporate Investment Grade Spreads & Crude Oil			95%	88%	63%	4%	0.12	0.29
	Terms of Trade & Retail Sales (6m ret)			96%	91%	79%	5%	0.09	0.17
				96%	90%	74%	5%	0.10	0.20
	Terms of Trade & Crude Oil			88%	83%	65%	12%	0.18	0.29
	Retail Sales (6m_ret) & Crude Oil			94% 99%	88%	70%	6%	0.12	0.24
		Channels: Money Supply & Industrial			97%	93%	2%	0.03	0.06
	Channels: Money Supp	ly & Refinancing		98%	93%	79%	3%	0.07	0.16
	Channels: Money Supp			96%	92%	72%	2%	0.09	0.21
	Channels: Money Supp			100%	96%	88%	1%	0.04	0.09
	Channels: Money Supp			95%	92%	72%	2%	0.09	0.21
	Channels: Industrial &			97%	91%	77%	4%	0.09	0.18
	Channels: Industrial &			99%	94%	88%	4%	0.06	0.10
Channels	Channels: Industrial &			96%	93%	84%	4%	0.07	0.13
	Channels: Industrial &			98%	93%	86%	5%	0.07	0.12
	Channels: Refinancing	& Competitiveness		98%	92%	79%	4%	0.08	0.17
	Channels: Refinancing	& Sentiment		95%	91%	74%	3%	0.09	0.20
	Channels: Refinancing			97%	91%	79%	5%	0.09	0.17
	Channels: Competitive			100%	98%	95%	1%	0.02	0.04
	Channels: Competitive			94%	87%	70%	8%	0.14	0.25
	Channels: Sentiment &	Commodities		100%	100%	100%	0%	0.00	0.00

# 18 months

18m Forecas	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive vs.				L = f * (Type2)	Error) + (1 - f) *
Channel	Signal Issued	A	В	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		
Citatillei	No Signal Issued	C	D					L= 0.25	L= 0.75
	Real M1 (12m_ret) & PI			92%	87%	78%	7%	0.11	0.19
		vestment Grade Spreads		93%	83%	73%	10%	0.14	0.23
	Real M1 (12m_ret) & Te			92%	79%	73%	18%	0.20	0.25
	Real M1 (12m_ret) & Re			91%	82%	73%	12%	0.16	0.23
	Real M1 (12m_ret) & Co			90%	81%	75%	16%	0.18	0.23
	PMI (12m_diff) & Inves			85%	85%	69%	6%	0.12	0.25
	PMI (12m_diff) & Term:			89%	79%	66%	13%	0.18	0.29
Single Indicators	PMI (12m_diff) & Retail			79%	79%	63%	10%	0.17	0.31
	PMI (12m_diff) & Crude			89%	81%	75%	16%	0.18	0.23
		Grade Spreads & Terms o		88%	82%	75%	14%	0.17	0.23
	Corporate Investment Grade Spreads & Retail Sales (6m_ret)			82%	85%	75%	8%	0.13	0.21
		Grade Spreads & Crude C	Dil	89%	77%	79%	24%	0.23	0.22
	Terms of Trade & Retail Sales (6m_ret)			86%	78%	70%	18%	0.21	0.27
	Terms of Trade & Crude Oil			83%	75%	64%	19%	0.23	0.31
	Retail Sales (6m_ret) &	Crude Oil		84%	74%	67%	21%	0.24	0.30
	Channels: Money Supp			91%	84%	75%	10%	0.14	0.22
	Channels: Money Supp			93%	85%	70%	6%	0.12	0.24
	Channels: Money Supp			92%	81%	78%	17%	0.18	0.21
	Channels: Money Supp	oly & Sentiment		90%	87%	76%	6%	0.11	0.20
	Channels: Money Supp	oly & Commodities		92%	79%	76%	19%	0.20	0.23
	Channels: Industrial & I			87%	86%	70%	5%	0.11	0.24
	Channels: Industrial &	Competitiveness		92%	81%	75%	16%	0.18	0.23
Channels	Channels: Industrial & 5	Sentiment		82%	80%	63%	9%	0.16	0.30
	Channels: Industrial &	Commodities		93%	85%	78%	10%	0.13	0.19
	Channels: Refinancing	& Competitiveness		93%	86%	76%	7%	0.12	0.20
	Channels: Refinancing	& Sentiment		83%	83%	66%	6%	0.13	0.27
	Channels: Refinancing	& Commodities		94%	88%	81%	7%	0.10	0.16
	Channels: Competitive			91%	79%	70%	15%	0.19	0.26
	Channels: Competitive			85%	75%	67%	20%	0.23	0.30
	Channels: Sentiment &			92%	82%	75%	13%	0.16	0.22

#### 24 months

24m Forecas	ting Horizon			AUROC	Correct Signal	<b>Crisis Detected</b>	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification Signal Issued	Recession A	No Recession B	True positive vs.	(A+D)/(A+B+C+D)	A/(A+C)	A / (B+D)	L = f * (Type2)	Error) + (1 - f) *
Channel	No Signal Issued	С	D	False Positive Rate	, ,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,,	. , ,	. , ,	L= 0.25	L= 0.75
	Real M1 (12m_ret) & P			80%	72%	76%	32%	0.30	0.26
		vestment Grade Spreads	5	85%	74%	67%	20%	0.23	0.30
	Real M1 (12m_ret) & Terms of Trade			86%	77%	77%	23%	0.23	0.23
	Real M1 (12m_ret) & R			82%	73%	71%	24%	0.26	0.28
	Real M1 (12m_ret) & C			85%	73%	67%	22%	0.25	0.30
	PMI (12m_diff) & Inves			79%	83%	73%	9%	0.13	0.22
	PMI (12m_diff) & Term			81%	66%	59%	28%	0.31	0.37
Single Indicators	PMI (12m_diff) & Retai			70%	69%	57%	21%	0.27	0.38
-	PMI (12m_diff) & Crude			82%	69%	62%	24%	0.28	0.35
	Corporate Investment Grade Spreads & Terms of Trade			85%	74%	71%	23%	0.25	0.28
		Grade Spreads & Retail S		79%	82%	73%	10%	0.14	0.22
	Corporate Investment Grade Spreads & Crude Oil			88%	76%	77%	26%	0.25	0.23
	Terms of Trade & Retail Sales (6m_ret)			83%	69%	62%	26%	0.29	0.35
	Terms of Trade & Crude Oil			82%	72%	68%	24%	0.26	0.30
	Retail Sales (6m_ret) &			82% 80%	72%	70%	27%	0.28	0.29
		Channels: Money Supply & Industrial			71%	73%	31%	0.30	0.28
	Channels: Money Supp			82%	79%	62%	7%	0.14	0.30
	Channels: Money Supp			86%	75%	75%	26%	0.25	0.25
	Channels: Money Supp			79%	72%	75%	31%	0.30	0.27
	Channels: Money Supp			87%	76%	75%	23%	0.24	0.25
	Channels: Industrial &			75%	77%	59%	8%	0.16	0.32
	Channels: Industrial &			86%	72%	68%	26%	0.27	0.30
Channels	Channels: Industrial &	Sentiment		70%	70%	62%	22%	0.26	0.34
	Channels: Industrial &	Commodities		87%	75%	73%	23%	0.24	0.26
	Channels: Refinancing			89%	74%	70%	22%	0.24	0.28
	Channels: Refinancing			83%	76%	58%	9%	0.17	0.34
	Channels: Refinancing	& Commodities		91%	76%	75%	22%	0.23	0.25
	Channels: Competitive	ness & Sentiment		86%	71%	68%	27%	0.28	0.30
	Channels: Competitive	ness & Commodities		84%	75%	76%	27%	0.26	0.25
	Channels: Sentiment &	Commodities		87%	74%	73%	26%	0.26	0.26

Table A14: Submodel 3 - Unstable Surges over Various Time Horizons 6 months

6m Forecasti	ng Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification Signal Issued No Signal Issued	Recession A C	No Recession B D	True positive rate vs. False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	L = f * (Type2Error) + L= 0.25	L= 0.75
	Real M1 & HP_Diff Lag			97%	93%	84%	5%	0.08	0.13
A A C b -	Real M1 & HP_Diff Lag	g (t-1yr)		100%	96%	93%	3%	0.04	0.06
Money Supply	Real M1 & HP_Diff Lag	(t-2yr)		94%	91%	70%	3%	0.10	0.23
	Single Indicator: R	leal M1 (12m_ret)		94%	91%	70%	2%	0.09	0.23
	ESTOXX50 (12m_ret) &			76%	80%	26%	4%	0.22	0.57
Coult Advantage	ESTOXX50 (12m_ret) &	HP_Diff Lag (t-1yr)		77%	78%	21%	6%	0.24	0.61
Equity Markets	ESTOXX50 (12m_ret) 8			78%	80%	33%	6%	0.22	0.52
	ESTOXX50 (12m_re			77%	78%	21%	6%	0.24	0.61
	Residential Building Pe	ermits (6m_ret) & HP_Dif	f Lag (t0)	86%	82%	47%	8%	0.19	0.42
D. Hallan Danielan	Residential Building Pe	ermits (6m_ret) & HP_Diff	Lag (t-1yr)	82%	84%	42%	3%	0.17	0.44
<b>Building Permits</b>	Residential Building Pe	ermits (6m_ret) & HP_Diff	Lag (t-2yr)	81%	84%	44%	4%	0.17	0.43
	Residential Buildir			78%	84%	40%	3%	0.17	0.46
	PMI (12m_diff) & HP_I	Diff Lag (t0)		97%	91%	81%	6%	0.09	0.15
	PMI (12m_diff) & HP_D	Diff Lag (t-1yr)		95%	90%	74%	5%	0.10	0.20
Sentiment	PMI (12m_diff) & HP_D	Diff Lag (t-2yr)		94%	92%	74%	3%	0.08	0.20
	Single Indicator: P			95%	90%	74%	5%	0.10	0.20
	HICP (12m_ret) & HP_D			86%	86%	37%	0%	0.16	0.47
Indian's	HICP (12m_ret) & HP_I			95%	91%	77%	4%	0.09	0.18
Inflation	HICP (12m_ret) & HP_D	Diff Lag (t-2yr)		86%	86%	47%	2%	0.15	0.41
	Single Indicator: H	IICP (12m_ret)		86%	86%	37%	0%	0.16	0.47
	Retail Sales (6m_ret) 8			94%	88%	74%	8%	0.13	0.21
D-4-11	Retail Sales (6m_ret) &			91%	87%	53%	3%	0.14	0.36
Retail	Retail Sales (6m_ret) &			91%	86%	49%	3%	0.15	0.39
		letail Sales (6m_ret)		90%	86%	51%	4%	0.15	0.38
		ector (12m_ret) & HP_Di		82%	84%	42%	3%	0.17	0.44
		ector (12m_ret) & HP_Diff		77%	77%	23%	7%	0.24	0.59
Credit Dynamics		ector (12m_ret) & HP_Diff		61%	79%	9%	0%	0.23	0.68
	Single Indicator: N	AFI Loans to Private Secto	r (12m_ret)	54%	77%	0%	0%	25%	75%
	Crude Oil & HP_Diff La	ng (t0)		89%	81%	53%	10%	0.19	0.37
Oil	Crude Oil & HP_Diff Lag			84%	76%	37%	12%	0.25	0.50
UII	Crude Oil & HP_Diff Lag			85%	81%	42%	8%	0.20	0.46
	Single Indicator: C			82%	80%	37%	8%	0.21	0.49
	3m EURIBOR (12m_diff			58%	77%	0%	0%	0.25	0.75
Central Bank	3m EURIBOR (12m_dif	ff) & HP_Diff Lag (t-1yr)		95%	89%	72%	6%	0.12	0.22
Policy	3m EURIBOR (12m_diff	f) & HP_Diff Lag (t-2yr)		54%	77%	0%	0%	0.25	0.75
· Only	Single Indicator: 3			62%	77%	0%	0%	0.25	0.75

#### 18 months

18m Forecas	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive rate vs.				L = f * (Type2Error) +	(1 - f) * (Type1Error)
	Signal Issued	A	В		(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)		
Channel	No Signal Issued	С	D	False Positive Rate				L= 0.25	L= 0.75
	Real M1 & HP_Diff Lag			94%	83%	79%	15%	0.16	0.19
	Real M1 & HP_Diff Lag	(t-1yr)		90%	85%	82%	14%	0.15	0.17
Money Supply	Real M1 & HP_Diff Lag	(t-2yr)		90%	77%	72%	20%	0.22	0.26
	Single Indicator: F	leal M1 (12m_ret)		90%	81%	75%	15%	0.17	0.23
	ESTOXX50 (12m_ret) 8			64%	62%	25%	16%	0.30	0.60
F	ESTOXX50 (12m_ret) 8	k HP_Diff Lag (t-1yr)		66%	65%	36%	17%	0.29	0.52
Equity Markets	ESTOXX50 (12m_ret) 8	HP_Diff Lag (t-2yr)		65%	61%	21%	14%	0.30	0.63
	ESTOXX50 (12m_r			65%	60%	24%	18%	0.32	0.61
		ermits (6m_ret) & HP_Di		91%	79%	61%	9%	0.17	0.31
D		ermits (6m_ret) & HP_Dif		74%	71%	51%	17%	0.25	0.41
<b>Building Permits</b>		ermits (6m_ret) & HP_Dif	Lag (t-2yr)	70%	69%	39%	13%	0.25	0.49
	Residential Buildi	ng Permits (6m_ret)		69%	67%	40%	16%	0.27	0.49
	PMI (12m_diff) & HP_			89%	84%	76%	11%	0.14	0.21
C+!	PMI (12m_diff) & HP_E			81%	76%	63%	16%	0.21	0.32
Sentiment	PMI (12m_diff) & HP_E			81%	75%	57%	13%	0.21	0.36
	Single Indicator: P			78%	74%	51%	12%	0.21	0.40
	HICP (12m_ret) & HP_I			75%	66%	52%	25%	0.31	0.42
Inflation	HICP (12m_ret) & HP_	Diff Lag (t-1yr)		75%	68%	52%	22%	0.29	0.41
inflation	HICP (12m_ret) & HP_I	Diff Lag (t-2yr)		75%	70%	52%	19%	0.27	0.41
	Single Indicator: F			75%	67%	52%	23%	0.29	0.42
	Retail Sales (6m_ret) 8			86%	83%	79%	14%	0.16	0.19
D-4-II	Retail Sales (6m_ret) 8			76%	73%	49%	12%	0.22	0.41
Retail	Retail Sales (6m_ret) 8			75%	72%	46%	12%	0.22	0.43
		letail Sales (6m_ret)		75%	71%	48%	14%	0.23	0.43
		ector (12m_ret) & HP_D		74%	70%	45%	14%	0.24	0.45
Condit Domestics		ector (12m_ret) & HP_Dif		62%	75%	42%	4%	0.17	0.45
Credit Dynamics		ector (12m_ret) & HP_Dif		61%	59%	13%	12%	0.31	0.68
		AFI Loans to Private Sect	or (12m_ret)	60%	68%	30%	8%	0.24	0.55
	Crude Oil & HP_Diff La	ng (t0)		88%	73%	64%	22%	0.26	0.32
Oil	Crude Oil & HP_Diff La	g (t-1yr)		80%	71%	58%	20%	0.26	0.36
UII	Crude Oil & HP_Diff La			80%	74%	61%	19%	0.24	0.34
	Single Indicator: C			79%	74%	61%	19%	0.24	0.34
	3m EURÍBOR (12m_dif			72%	77%	46%	4%	0.16	0.41
Central Bank		tt) & HP_Diff Lag (t-1yr)		72%	81%	55%	4%	0.14	0.35
Policy	3m EURIBOR (12m_dif	f) & HP_Diff Lag (t-2yr)		66%	74%	36%	2%	0.17	0.49
	Single Indicator: 3	m FURIBOR		65%	69%	39%	13%	0.25	0.49

# 24 months

24m Forecas	ting Horizon			AUROC	Correct Signal	Crisis Detected	False Alarm	Risk-Affine	Risk-Averse
Transmission	Classification	Recession	No Recession	True positive rate vs.				L = f * (Type2Error)	(1 - f) * (Type1Error
Channel	Signal Issued No Signal Issued	A	B D	False Positive Rate	(A+D)/(A+B+C+D)	A / ( A + C)	A / (B+D)	L= 0.25	L= 0.75
	Real M1 & HP Diff Lag	z (t0)		88%	79%	82%	24%	0.23	0.19
	Real M1 & HP_Diff Lag	(t-1vr)		82%	73%	77%	31%	0.29	0.25
Money Supply	Real M1 & HP Diff Lag	(t-2vr)		81%	73%	77%	31%	0.29	0.25
	Single Indicator: R			81%	73%	77%	30%	0.28	0.25
	ESTOXX50 (12m_ret) 8	k HP Diff Lag (t0)		61%	56%	33%	24%	0.35	0.56
	ESTOXX50 (12m_ret) &			57%	59%	38%	23%	0.33	0.52
Equity Markets	ESTOXX50 (12m ret) &	HP Diff Lag (t-2vr)		58%	57%	35%	24%	0.34	0.55
	ESTOXX50 (12m r	et)		58%	57%	34%	23%	0.34	0.55
	Residential Building Pe	ermits (6m ret) & HP Di	ff Lag (t0)	92%	86%	77%	7%	0.11	0.19
	Residential Building Pe	ermits (6m_ret) & HP_Diff	Lag (t-1yr)	69%	64%	53%	27%	0.32	0.42
Building Permits	Residential Building Pe	ermits (6m ret) & HP Diff	Lag (t-2vr)	68%	62%	52%	29%	0.34	0.43
	Residential Buildir	ng Permits (6m ret)	M.L	67%	62%	48%	27%	0.33	0.46
	PMI (12m diff) & HP I	Diff Lag (t0)		89%	77%	78%	24%	0.24	0.22
	PMI (12m diff) & HP D	Diff Lag (t-1yr)		64%	66%	53%	23%	0.29	0.41
Sentiment	PMI (12m diff) & HP D	Diff Lag (t-2yr)		63%	64%	48%	21%	0.29	0.44
	Single Indicator: P			66%	66%	52%	22%	0.29	0.42
	HICP (12m_ret) & HP_D			71%	57%	72%	57%	0.49	0.35
	HICP (12m_ret) & HP_I	Diff Lag (t-1yr)		71%	57%	62%	48%	0.45	0.40
Inflation	HICP (12m ret) & HP D	Diff Lag (t-2yr)		71%	56%	59%	47%	0.45	0.42
	Single Indicator: H	HCP (12m ret)		71%	56%	58%	46%	0.45	0.43
	Retail Sales (6m_ret) 8			88%	83%	80%	14%	0.16	0.19
	Retail Sales (6m_ret) &	HP_Diff Lag (t-1yr)		71%	69%	59%	22%	0.27	0.36
Retail	Retail Sales (6m_ret) &			71%	68%	58%	23%	0.28	0.37
		tetail Sales (6m_ret)		71%	69%	58%	22%	0.27	0.37
	MFI Loans to Private S	ector (12m_ret) & HP_Di	ff Lag (t0)	70%	66%	58%	28%	0.31	0.38
Consider Dominion	MFI Loans to Private Se	ector (12m_ret) & HP_Dif	Lag (t-1yr)	64%	60%	48%	29%	0.35	0.46
redit Dynamics	MFI Loans to Private Se	ector (12m_ret) & HP_Dif	Lag (t-2yr)	60%	57%	43%	31%	0.38	0.50
	Single Indicator: N	AFI Loans to Private Secto	r (12m_ret)	61%	57%	43%	30%	0.37	0.50
	Crude Oil & HP_Diff La	ng (t0)		86%	80%	85%	23%	0.21	0.17
Oil	Crude Oil & HP_Diff La	g (t-1yr)		79%	75%	72%	22%	0.24	0.26
Oil	Crude Oil & HP_Diff La	g (t-2yr)		79%	73%	67%	22%	0.25	0.30
	Single Indicator: C			79%	75%	71%	22%	0.24	0.27
	3m EURIBOR (12m_dif			70%	75%	65%	16%	0.21	0.30
Central Bank	3m EURIBOR (12m_diff	f) & HP_Diff Lag (t-1yr)		66%	72%	58%	16%	0.22	0.35
Policy	3m EURIBOR (12m_diff			68%	69%	61%	23%	0.27	0.35
	Single Indicator: 3	m EURIBOR		63%	59%	42%	27%	0.35	0.50