



EUROPEAN CENTRAL BANK

EUROSYSTEM

Workshop on using big data for forecasting and statistics

Monday, 7 and Tuesday, 8 April 2014
European Central Bank, Eurotower
Frankfurt am Main





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CAN INFORMATION DEMAND HELP TO PREDICT STOCK MARKET LIQUIDITY ? GOOGLE IT!

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THE STORY

Theory

Infinite information processing abilities

New information is immediately incorporated into stock prices (Fama, 1970)

Merton (1987), Barber and Odeon (2008)

Demand for Information

Stock market dynamics

Reality

Infobesity: Huge amounts of information

Limited time and scarce attention



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Introduction Contribution Data & Methodology Empirical Results Conclusion

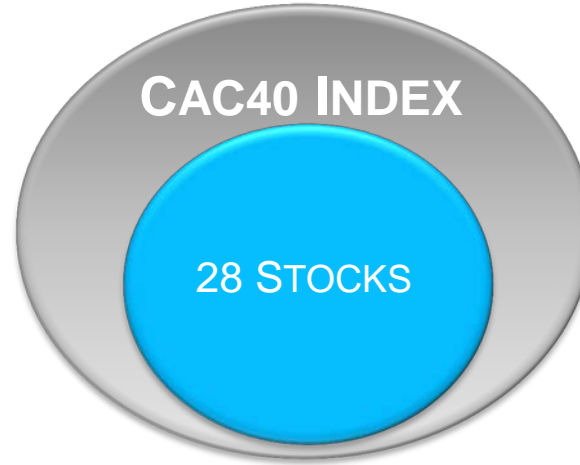
OUR AIM

EXISTING STUDIES

- ▶ Da et al. (2011), Vlastakis and Markellos (2012), Aouadi et al.(2013)
- ▶ Drake et al. (2012) and Vlastakis and Markellos (2012)
- ▶ Askitas and Zimmermann (2009), Kulkarni et al. (2009), Da et al. (2011) and Zielinski (2011)

CENTRAL QUESTION

Can **information demand** help to forecast the French stock market **liquidity**?



1 SAMPLE



2 DATA



Google Insights for search data, Several liquidity measures

3 PERIOD



From *January 09, 2004* to *June, 22, 2012*



MODELS

▶ IN SAMPLE ANALYSIS

1 $Turnover_{it} = \alpha + \beta_1 Turnover_{i,t-1} + \beta_2 Ln(Number_of_Analysts)_{i,t-1} + \beta_3 Ln(Number_of_Employees)_{i,t-1} + \beta_4 Ln(Market_Value)_{i,t-1} + \beta_5 Inverse_of_Stock_Price_{i,t-1} + \beta_6 Absolute_return_{i,t-1} + \beta_7 Std_Dev_{i,t-1} + \varepsilon_{it}$

2 $Turnover_{it} = \alpha + \beta_1 Turnover_{i,t-1} + \beta_2 Ln(Number_of_Analysts)_{i,t-1} + \beta_3 Ln(Number_of_Employees)_{i,t-1} + \beta_4 Ln(Market_Value)_{i,t-1} + \beta_5 Inverse_of_Stock_Price_{i,t-1} + \beta_6 Absolute_return_{i,t-1} + \beta_7 Std_Dev_{i,t-1} + \lambda_1 Ln(GSV)_{i,t-1} + \lambda_2 Ln(GSV)_{CAC40,t-1} + \varepsilon_{it}$

▶ OUT-OF-SAMPLE ANALYSIS

1 $RMSE_i = \sqrt{h^{-1} \sum_{t=T+1}^{T+h} (l_{it} - \hat{l}_{it})^2}$

2 $MAPE_i = h^{-1} \sum_{t=T+1}^{T+h} \left| \frac{l_{it} - \hat{l}_{it}}{l_{it}} \right|$



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NEW RESULTS

- 1 Google variables and turnover are **positively** and **strongly** correlated in most cases
- 2 Google search variables contribute to **better** understand liquidity formation in the French stock market
- 3 Higher level of Google search volume leads to **higher** turnover for the majority of our sample
- 4 Model (2) leads to **more precise** out-of-sample forecasts in most cases and for almost all horizons



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SUMMARY & CONCLUSION

1 Implications for Liquidity Forecasting

- ▶ It is possible to predict liquidity using publicly available variables
- ▶ Information demand is able to refine liquidity forecasting results

2 Implications for Academics & Practitioners

- ▶ Better understand the dynamic of liquidity series, and help portfolio managers to conceive less costly trading strategies.
- ▶ The importance of including online investor search behavior in forecasting important economic outcomes.