Off to the Races: A Comparison of Machine Learning and Alternative Data for Predicting Economic Indicators by J. Chen, A. Dunn, K. Hood, A. Driessen, and A. Batch

> Discussion by Francis X. Diebold University of Pennsylvania

CRIW 2019, Bethesda

March 14, 2019



Econometrics / Statistics Perspectives

ML: $\hat{y}(=x'\hat{\beta})$ ("prediction")

Time-series:

Econometrics: \hat{y} ("prediction") (Time series econometrics \leftrightarrow predictive dynamic econometric modeling)

ML: \hat{y} ("prediction") (???)

So what's new in ML?



Time Series Econometrics (TSE) vs. ML

Significant TSE / ML overlap:

- Acknowledge misspecification throughout
- Seek good out-of-sample predcitive approximations
 - Use the relevant loss function
 - Shrinkage
 - Selection
 - Forecast combination ("ensemble averaging")

ML goes farther in some important directions:

- High dimensionality
 - Nonlinearity
- Interesting new procedures



Time Series Econometrics (TSE) vs. ML

But TSE Goes Much Farther in Important Macroeconometric Directions...

– Trend

- Seasonality

- Serial correlation & cycles

- Workhorse linear models (VAR, ...)

- Summarizing voluminous results

(Impulse-response fns, variance decomps, Granger causality, ...)

- Customized reduced-rank linear models (DFM, FAVAR, ECM, ...)
 - Customized nonlinear models (regime-switching, volatility)
 - Structural evolution and breaks
 - Quantifying forecast uncertainty



In the Trenches, Down and Dirty...

$GDP \supset CE \supset PCE \supset PCE \supset PCEs_i$

This paper is interested in PCES.

PCES is partly based on the Quarterly Survey of Services (QSS). (The *PCES*_i are informed by the *QSS*_j only from release 3 onward.)

One would like to make the QSS more timely, by nowcasting.

Do ML nowcasting "regressions" of QSS components on timely x's: $QSS_{it} \rightarrow x_{1t}, ..., x_{Kt}, i = 1, ..., 188$

x's include both BLS data (from CES and CPI) and private data (First Data credit cards and Google Trends)



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- 10. ML emphasis on ensemble averaging probably *is* highly relevant (e.g., random forrests)

