

Out-of-Sample Equity Premium Prediction: Combination Forecasts and Links to the Real Economy

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Overview

- Introduction
- Model
- Main Results
- Connection with Real Economy
- Conclusion

Introduction

Background & Problems

- Numerous economic variables have been proposed as predictors of stock returns, and in-sample tests shows that there is significant evidence to return predictability.
- However, Welch and Goyal (2008) show that a long list of predictors from the literature is unable to deliver consistent superior out-of-sample forecast relative to historical average.

Solution

- A combination approach to the out-of-sample equity premium forecasting problem.

Model

Predictive Regression Model

$$r_{t+1} = \alpha_j + \beta_j x_{j,t} + \epsilon_{t+1}$$

- As in Welch and Goyal (2008), we generate out-of-sample forecasts using a recursive (expanding) estimation window.
- Followed by Campbell and Thompson (2008) and Welch and Goyal (2008), the historical average equity premium serves as a benchmark forecasting model.

Forecast Combination

Forecast Combination

$$\hat{r}_{c,t+1} = \sum_{i=1}^N \omega_{i,t} \hat{r}_{i,t+1}$$

- We use two classes of simple averaging schemes:
 - Simple average scheme: mean, median, trimmed mean.
 - Stock and Watson (2004): DMSPE.

DMSPE

Weights of DMSPE

$$\omega_{i,t} = \phi_{i,t}^{-1} / \sum_{j=1}^N \phi_{j,t}^{-1}$$
$$\phi_{i,t} = \sum_{s=m}^{t-1} \theta^{t-1-s} (r_{s+1} - \hat{r}_{i,s+1})^2$$

- Assigns greater weights to model forecasts that have lower MSPE.
- When $\theta < 1$, we attach greater weights to the recent forecast accuracy.
- We consider 1.0 and 0.9 for θ .

Forecast Evaluation

- Two methods are used: Out-of-Sample R^2 , realized utility gains.

Out-of-sample R^2 : R_{OS}^2 , by Campbell and Thompson (2008)

$$R_{OS}^2 = 1 - \frac{\sum_{k=q_0+1}^q (r_{m+k} - \hat{r}_{m+k})^2}{\sum_{k=q_0+1}^q (r_{m+k} - \bar{r}_{m+k})^2}$$

$$f_{t+1} = (r_{t+1} - \bar{r}_{t+1})^2 - [(r_{t+1} - \hat{r}_{t+1})^2 - (\bar{r}_{t+1} - \hat{r}_{t+1})^2]$$

- When $R_{OS}^2 > 0$, forecast outperforms the historical average forecast.
- f_{t+1} is MSPE-adjusted statistic, by regressing it on a constant and compute the t-statistics, we can obtain p-value, which is used to test whether $R_{OS}^2 > 0$ is significant.

Forecast Evaluation

- R_{OS}^2 measure has one limitation: it does not explicitly account for the risk borne by an investor over the out-of-sample period.
- **Solution:** Realized utility gains for a mean-variance investor.

Utility Function

$$w_{0,t} = \frac{1}{\gamma} \left(\frac{\bar{r}_{t+1}}{\hat{\sigma}_{t+1}^2} \right)$$

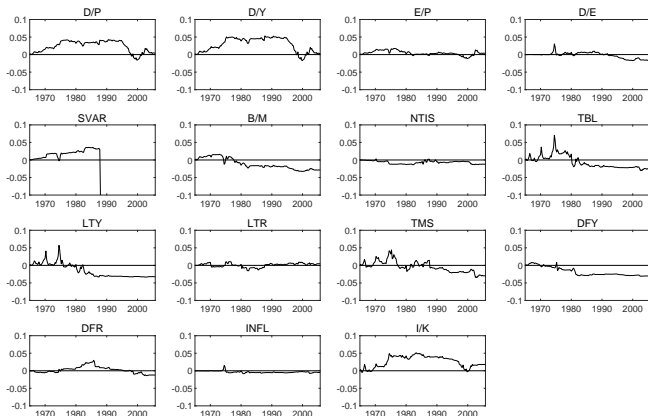
$$\hat{v}_0 = \hat{\mu}_0 - \left(\frac{1}{2} \right) \gamma \hat{\sigma}_0^2$$

$$w_{j,t} = \frac{1}{\gamma} \left(\frac{\hat{r}_{t+1}}{\hat{\sigma}_{t+1}^2} \right)$$

$$\hat{v}_j = \hat{\mu}_j - \left(\frac{1}{2} \right) \gamma \hat{\sigma}_j^2$$

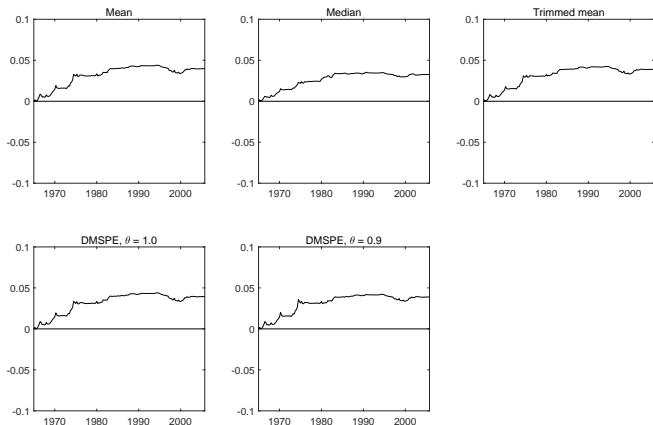
- We report the difference in utility to express the percentage return.

Out-of-Sample Forecasting Results



- If the curve increases, the regressive model outperforms.
- If the curve is higher at the end, our model has a lower MSPE.
- **Difficult to identify a predictor that reliably outperforms.**

Out-of-Sample Forecasting Results



- The combination forecasts delivers more consistent out-of-sample gains.

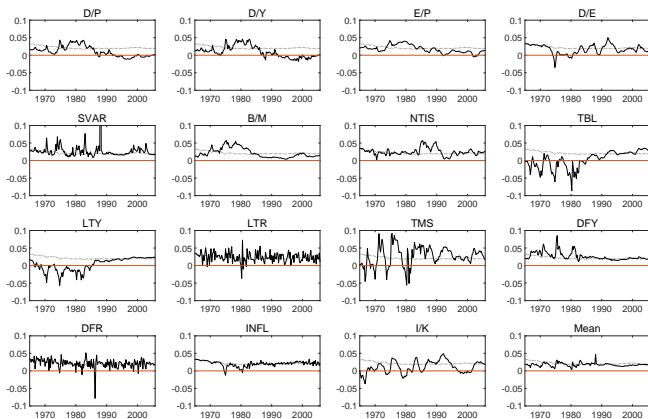
Out-of-Sample Forecasting Results

Table 1
Equity premium out-of-sample forecasting results for individual forecasts and combining methods

Individual predictive regression model forecasts						Combination forecasts		
Predictor	R_{OIS}^2 (%)	Δ (%)	Predictor	R_{OIS}^2 (%)	Δ (%)	Combining method	R_{OIS}^2 (%)	Δ (%)
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Panel A. 1965:1–2005:4 out-of-sample period								
<i>D/P</i>	0.34*	0.55	<i>LTY</i>	-3.09	2.29	Mean	3.58***	2.34
<i>D/Y</i>	0.25*	1.41	<i>LTR</i>	0.33	1.30	Median	3.04***	1.03
<i>E/P</i>	0.36	0.64	<i>TMS</i>	-2.96	5.14	Trimmed mean	3.51***	2.11
<i>D/E</i>	-1.42	0.58	<i>DFY</i>	-2.72	-0.83	DMSPE, $\theta = 1.0$	3.54***	2.41
<i>SVAR</i>	-12.97	0.13	<i>DFR</i>	-1.10	0.57	DMSPE, $\theta = 0.9$	3.49***	2.59
<i>B/M</i>	-2.60	-0.58	<i>INFL</i>	-0.84	1.39			
<i>NTIS</i>	-0.91	0.08	<i>I/K</i>	1.44**	2.80	Mean, CT	3.23***	1.25
<i>TBL</i>	-2.78	2.60						
Panel B. 1976:1–2005:4 out-of-sample period								
<i>D/P</i>	-5.08	-0.70	<i>LTY</i>	-5.59	-0.89	Mean	1.19*	0.57
<i>D/Y</i>	-6.22	-0.54	<i>LTR</i>	-0.27	1.43	Median	1.51**	0.53
<i>E/P</i>	-1.70	0.75	<i>TMS</i>	-7.24	2.08	Trimmed mean	1.23*	0.59
<i>D/E</i>	-2.26	-1.65	<i>DFY</i>	-2.48	-1.18	DMSPE, $\theta = 1.0$	1.11*	0.54
<i>SVAR</i>	-22.47	0.06	<i>DFR</i>	-2.14	-0.64	DMSPE, $\theta = 0.9$	1.01*	0.46
<i>B/M</i>	-4.72	-1.27	<i>INFL</i>	-0.08	0.45			
<i>NTIS</i>	0.10	0.60	<i>I/K</i>	-3.47	-0.85	Mean, CT	1.20*	0.55
<i>TBL</i>	-7.31	-0.82						
Panel C. 2000:1–2005:4 out-of-sample period								
<i>D/P</i>	10.32*	12.96	<i>LTY</i>	-0.32	0.24	Mean	3.04**	2.31
<i>D/Y</i>	10.40*	12.98	<i>LTR</i>	-1.72	2.57	Median	1.56*	0.28
<i>E/P</i>	8.02*	9.53	<i>TMS</i>	-4.98	4.23	Trimmed mean	2.98**	2.12
<i>D/E</i>	0.56	0.50	<i>DFY</i>	-0.53	-1.52	DMSPE, $\theta = 1.0$	2.56**	1.65
<i>SVAR</i>	-5.62	-1.64	<i>DFR</i>	-2.10	1.76	DMSPE, $\theta = 0.9$	2.66**	1.97
<i>B/M</i>	2.32	3.09	<i>INFL</i>	-1.42	0.57			
<i>NTIS</i>	-4.09	1.33	<i>I/K</i>	8.96**	9.13	Mean, CT	2.43**	1.32
<i>TBL</i>	-2.50	-0.20						

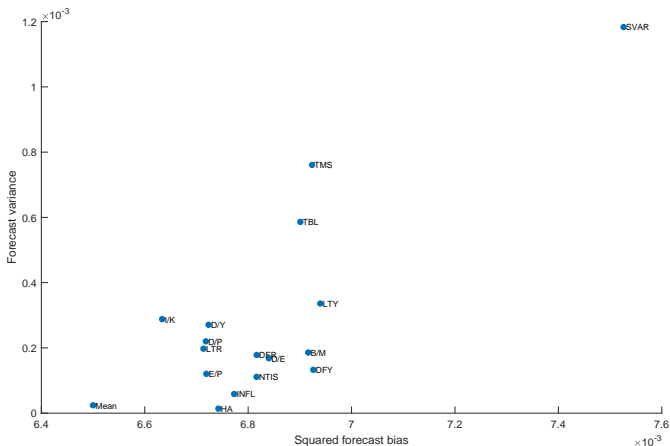
- Out-of-sample predictive ability deteriorates in 1976 - 2005.
- **DMSPE combination selects weights relatively close to $1/N$.**

Forecast Stabilization



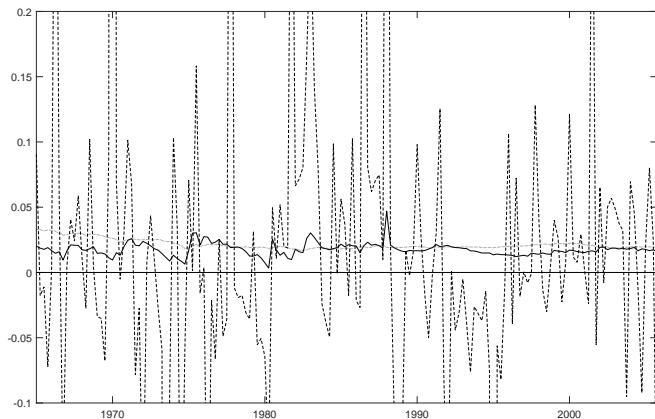
- The forecast combination reduces forecast variability.

Forecast Stabilization



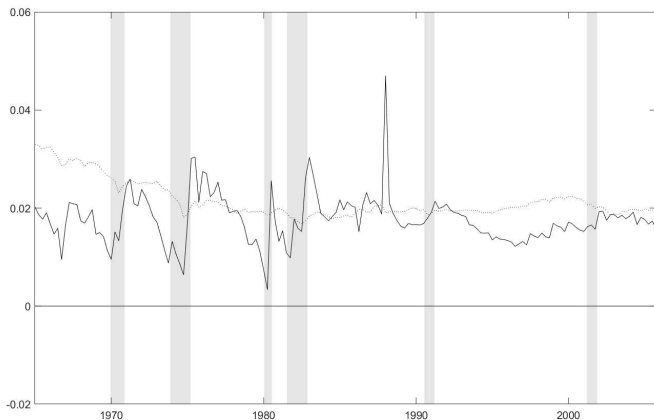
- The mean combination forecast has a variance moderately higher than that of the historical average forecast, and has a square bias substantially below the historical average.

Comparison to Kitchen-Sink Model



- **Unrestricted kitchen sink model forecast is much more volatile, even more volatile than individual predictive regression model forecasts.**

Business Cycle Phases



- **Figure 3 & Figure 6 shows that combination approach includes relevant macroeconomic information missed by historical**

Forecast Gains During "Good" and "Bad" Periods

Table 5

R_{OS}^2 statistics for out-of-sample equity premium combination forecasts during good, normal, and bad growth periods, 1965:1–2005:4

Combining method	Forecast horizon: one quarter				Forecast horizon: four quarters			
	Overall	Good	Normal	Bad	Overall	Good	Normal	Bad
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)	(9)
Panel A. Sorting on real GDP growth								
Mean	3.58***	1.82	1.71	6.17***	8.19***	3.07	3.63*	11.58***
Median	3.04***	2.67**	0.39	5.02***	6.99***	12.74***	6.35**	5.23***
Trimmed mean	3.51***	2.25*	1.24	5.94***	8.13***	5.41*	4.01*	10.63***
DMSPE, $\theta = 1.0$	3.54***	1.71	1.56	6.26***	7.87***	2.32	3.15	11.46***
DMSPE, $\theta = 0.9$	3.49***	1.60	1.36	6.33***	5.96***	4.71*	0.27	8.27***
Panel B. Sorting on real profit growth								
Mean	3.58***	2.87*	-1.03	7.94***	8.19***	0.93	4.89*	14.72***
Median	3.04***	2.56**	0.21	5.74***	6.99***	1.14	8.00**	10.18***
Trimmed mean	3.51***	2.85*	-0.67	7.47***	8.13***	1.74	5.83**	13.55***
DMSPE, $\theta = 1.0$	3.54***	2.74*	-1.21	8.08***	7.87***	0.16	4.41	14.78***
DMSPE, $\theta = 0.9$	3.49***	2.51	-1.56	8.40***	5.96***	-4.28	2.00	14.70***
Panel C. Sorting on real net cash flow growth								
Mean	3.58***	5.44**	2.17*	4.63**	8.19***	3.29*	8.81***	11.42***
Median	3.04***	4.12***	1.80**	4.25**	6.99***	4.99***	6.17**	9.48***
Trimmed mean	3.51***	5.01**	2.36**	4.47**	8.13***	4.39**	9.13***	10.04***
DMSPE, $\theta = 1.0$	3.54***	5.51**	2.13*	4.52**	7.87***	2.97*	8.50***	11.09***
DMSPE, $\theta = 0.9$	3.49***	5.88**	1.84*	4.15*	5.96***	0.53	6.66**	9.56***

- Out-of-sample gains for the combination forecasts are concentrated in especially "bad" periods.

Conclusion

- Forecast combination appears successful for out-of-sample equity premium prediction, avoiding being too volatile as in individual forecast model, or being too smooth as in historical average model.
- Forecast of equity premium are linked to the real economy.
- Forecasting combining methods stems from highly uncertain, complex and constantly evolving data-generating process underlying expected equity returns.

Thanks!