Does
Academic
Research
Destroy
Stock
Return Predictability?

David McLean and Jeffrey Pontiff

Does Academic Research Destroy Stock Return Predictability?

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Question

What exactly happens to return predictability outside of a study's sample period?

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What could happen?

- Perhaps the findings are spurious and only fit the partitioned sample period
- 2 Or maybe there will be no change, and return predictability will persist
- 3 Or return predictability will diminish post-publication, suggesting mispricing

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Hypotheses

- If findings are spurious, there should be no returns outside of the sample period
- 2 If return predictability reflects rational expectations, findings should remain
- 3 If return predictability is due to mispricing, then return predictability should disappear or be decay after a paper is published

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Key Findings

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David McLean and Jeffrey Pontiff Using 97 predictors from 79 studies the average predictor long-short quintile portfolio return:

- 1 Declines 26% out-of-sample
- 2 Shrinks 58% post-publication

Additionally, decay is larger for:

- 3 Predictor portfolios with larger in-sample returns and higher t-statistics
- 4 Predictors constructed from price and trading data

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Contribution

- Academic publications transmit information to sophisticated investors
- Or we do something worthwhile and important!

Research Method

Does
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- Exclude time series predictability
- Focus on studies in peer-reviewed finance/economic/accounting literature

Research Method

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- Gather 97 cross-sectional relations from 79 studies
- Form long-short quintiles based on results

Some Concerns

- Data no longer available to construct some measures get proxies
- Rate of increase/decrease form long-short portfolios with extreme 20 percentiles
- Dummy variables separate into long or short side of the portfolio
- Discrete values follow original research

Summary Statistics

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Table I Summary Statistics

This table reports summary statistics for the predictor portfolios studied in this paper. The returns are equal-weighted by predictor portfolio, that is, we first estimate the statistic for each predictor portfolio, and then take an equal-weighted average across the portfolios. The reported standard deviations are the standard deviations of the predictors' mean returns. Our sample period ends in 2013.

Number of predictor portfolios	97
Predictors portfolios with t -statistic > 1.5	85 (88%)
Mean publication year	2000
Median publication year	2001
Predictors from finance journals	68 (70%)
Predictors from accounting journals	27 (28%)
Predictors from economics journals	2 (2%)
Mean portfolio return in-sample	0.582
Standard deviation of mean in-sample portfolio return	0.395
Mean observations in-sample	323
Mean portfolio return out-of sample	0.402
Standard deviation of mean out-of-sample portfolio return	0.651
Mean observations out-of-sample	56
Mean portfolio return post-publication	0.264
Standard deviation of mean post-publication portfolio return	0.516
Mean observations post-publication	156

Empirical Analysis

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Baseline Regression:

 $R_{it} = \alpha_i + \beta_1 \ Post \ Sampe \ Dummy_{i,t} + \beta_2 \ Post \ Publication \ Dummy_{i,t} + e_{it}$

Correlation

- Overall correlation is 0.033, but some will be higher
- Compute standard errors using feasible generalized least squares (FGLS)

Empirical Analysis

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Formal Hypotheses

- If statistical biases are the source of in-sample predictability, then the coefficients on both the post-sample and post-publication dummies should be -0.582
- 2 If predictors' returns are entirely the result of mmispricing and arbitrage resulting from publication corrects all mispricing, the the post-publication coefficient should be equal to -0.582 and the post-sample dummy should not be close to zero
- If there are no statistical biases and academic papers have no influence on investors' actions, then both of the coefficients should equal zero.

Regression of Predictor Portfolio Returns on Post-Sample and Post-Publication Indicators

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Variables	(1)	(2)
Post-Sample (S)	-0.150***	-0.180**
	(0.077)	(0.085)
Post-Publication (P)	-0.337***	-0.387***
	(0.090)	(0.097)
$S \times Mean$		
P × Mean		
$S \times t$ -statistic		
$P \times t$ -statistic		
Predictor FE?	Yes	Yes
Observations	51,851	45,465
Predictors (N)	97	85
Null: S = P	0.024	0.021
Null: $P = -1 \times (mean)$	0.000	0.000
Null: $S = -1 \times (mean)$	0.000	0.000

Regression of Predictor Portfolio Returns on Post-Sample and Post-Publication Indicators

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Question

Do predictor portfolio returns with higer in-sample means decline more post-publication?

Test

- Interact in-sample means with predictors
- Interaction for post sample is -0.532 and post-sample is 0.157
- Notice $0.157 + (-0.532 \times 0.582) = -0.153$ nearly the same as column (1)
- Indicates that predictor with larger return could have larger bias
- Or could indicate arbitrageurs are more likely to learn about higher returns before publication

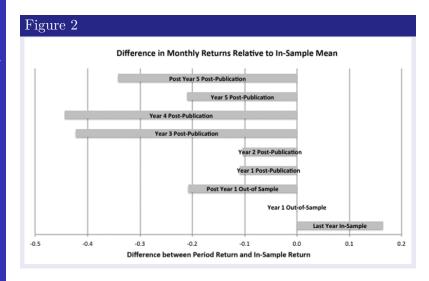
Regression of Predictor Portfolio Returns on Post-Sample and Post-Publication Indicators

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Variables	(1)	(2)	(3)	(4)
Post-Sample (S)	-0.150***	-0.180**	0.157	0.067
	(0.077)	(0.085)	(0.103)	(0.112)
Post-Publication (P)	-0.337***	-0.387***	-0.002	-0.120
	(0.090)	(0.097)	(0.078)	(0.114)
S × Mean			-0.532***	
			(0.221)	
P × Mean			-0.548***	
			(0.178)	
$S \times t$ -statistic				-0.061**
				(0.023)
$P \times t$ -statistic				-0.063**
				(0.018)
Predictor FE?	Yes	Yes	Yes	Yes
Observations	51,851	45,465	51,851	51,944
Predictors (N)	97	85	97	97
Null: S = P	0.024	0.021		
Null: $P = -1 \times (mean)$	0.000	0.000		
Null: $S = -1 \times (mean)$	0.000	0.000		

Result

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Controlling for Time Trends and Persistence

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Concern

 Perhaps the authors are measuring a time trend that proxies for lower costs of corrective trading

Control

■ Use time variable that is equal to 1/100 in January 1926 and increases by 1/100 each month in the sample.

Result

■ Time variable has significant and negative slope - suggests portfolio returns have declined overtime.

Controlling for Time Trends and Persistence

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Table III Time Trends and Persistence in Predictor Returns

The regressions reported in this table test for time trends and persistence in predictor returns. Post-Sample (S) is equal to one if the month is after the sample period used in the original study and zero otherwise. Post-Publication (P) is equal to one if the month is after the official publication date and zero otherwise. Time is the number of months divided by 100 post-January 1926. Post-1993 is equal to one if the year is greater than 1993 and zero otherwise. I-Month Return and 12-Month Return are the predictor's return from the last month and the cumulative return over the last 12 months. Standard errors (in parentheses) are computed under the assumption of contemporaneous cross-sectional correlation between panel portfolio residuals. *, **, and *** denote statistical significance at the 10%, 5%, and 1% levels, respectively.

Variable	(1)	(2)	(3)	(4)	(5)	(6)
Time	-0.069***		-0.069***			
	(0.011)		(0.026)			
Post-1993		-0.120	0.303***			
		(0.074)	(0.118)			
Post-Sample			-0.190**	-0.179**	-0.132*	-0.128
			(0.081)	(0.080)	(0.076)	(0.078)
Post-Publication			-0.362***	-0.310**	-0.295***	-0.258***
			(0.124)	(0.122)	(0.089)	(0.093)
1-Month Return					0.114***	
					(0.015)	
12-Month Return						0.020***
						(0.004)
Observations	51,851	51,851	51,851	51,851	51,754	50,687
Char. FE?	Yes	Yes	Yes	Yes	Yes	Yes
Time FE?	No	No	No	Yes	No	No

Do Predictor Types Vary?

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Four Predictor Types

- Event share issues, changes in analyst recommendations, R&D changes
- 2 Market volume, price, returns, shares outstanding, momentum
- 3 Valuation sales-to-price, book-to-maret
- 4 Fundamentals Debt, taxes, accruals

Do Predictor Types Vary?

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Variable	(1)	(2)	(3)	(4)
Post-Publication (P)	-0.208***	-0.316***	-0.310***	-0.301***
	(0.059)	(0.097)	(0.080)	(0.089)
Market	0.304***			
	(0.079)			
$P \times Market$	-0.244			
	(0.169)			
Event		-0.098**		
		(0.046)		
$P \times Event$		0.105		
		(0.091)		
Valuation			-0.056	
			(0.063)	
P × Valuation			0.186	
			(0.131)	
Fundamental				-0.201***
				(0.045)
$P \times Fundamental$				0.025
				(0.089)
Constant	0.482***	0.606***	0.585***	0.630***
	(0.036)	(0.052)	(0.000)	(0.053)
Observations	51,851	51,851	51,851	51,851
Predictors	97	97	97	97
$Type + (P \times Type)$	0.060	0.007	0.121	-0.176
p-value	0.210	0.922	0.256	0.012

Costly Arbitrage

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Here we look at averages, but this could highlight a few important factors and diminish others