

Give me a break: What does the equity premium compensate for?

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Motivation

Motivation I

Unresolved puzzle on the overall intertemporal risk-return relation

• Significant positive relation

(e.g. Harvey 1989, Bali/Peng 2006, Hedegaard/Hodrick 2016 among many others).

Weak or negative relation

(e.g. Campbell 1987, Glosten et al. 1993, Hedegaard/Hodrick 2014 among many others).

What might be reasons for the controversial results?

- Specification of volatility (e.g. Guo/Neely 2008, Leon et al. 2007).
- Omission of Merton's (1973) hedging component (e.g. Sruggs 1998).
- Others, such as irrational investor behavior (Yu/Yuan 2011, Nagel/Xu 2022, Lochstoer/Muir 2022).

Motivation II

Why do trading breaks matter?

- Empirics:
 - Market volatility is lower during periodic trading breaks (e.g. French/Roll 1980, Stoll/Whaley 1990, Kelly/Clark 2011).
 - Significant contribution to the overall equity premium (e.g. Cliff et al. 2008, Perras/Wagner 2020, Hendershott et al. 2020, Boyarchenko et al. 2023).
- Controversial theories:
 - Investors are averse of holding stocks over non-trading periods due to the lack of marketability (Longstaff 1995) and 'clientele effects' (Lou et al. 2019, Bogusslavsky 2021, Akbas et al. 2022).
 - Decrease in information asymmetry over the trading day increases the stock price resulting in a lower expected return over trading breaks (Hong/Wang 2000).

Contribution & Key Findings

Contribution & Key Findings

How does this study contribute to literature?

- \Rightarrow Evidence on the risk-return relation considering trading breaks.
- \Rightarrow The impact of trading breaks on the market risk premium (MRP).
- \Rightarrow Distinct pricing of trading and non-trading conditional volatility/illiquidity.

Main empirical findings:

- 1. Positive relation between the overall MRP and conditional risk and illiquidity during market closures.
- 2. Conditional risk and illiquidity during trading hours play a minor role in explaining the MRP.
- 3. Lack of market functionality and liquidity during closures results in a premium demanded by investors.
- 4. Analysis of intraday quotes relates the trading break premium to a decline before the official market close.

Model Setting

Model Setting

Assume a price process $\{P_t\}_{t\geq 0}$ with jumps at deterministic points in time:

$$\frac{dP_t}{P_{t-}} = \mu dt + \sigma dB_t + dX_t$$
where $X_t = \sum_{j=1}^{N_t} (e^{r_{co,j}} - 1),$
(1)

 $\{B_t\}_{t\geq 0}$ is a standard Brownian motion, dX_t is the discontinuous part of the process, N_t is a deterministic process counting the number of scheduled market closures at $\{\tau_j\}_{j\geq 1}$, up to time t.

Asset prices hence are:

$$P_t = P_0 \exp\left[\left(\mu - \frac{1}{2}\sigma^2\right)t + \sigma B_t + \sum_{\tau_j \le t} r_{co,j}\right].$$
(2)

Intertemporal Pricing Equation

"In equilibrium, investors are compensated in terms of expected return, for bearing (systematic) market risk, and for bearing the risk of unfavorable shifts in the investment opportunity set."

- Merton (1973, p. 882)

Our basic model at the aggregate market level is given by

$$E_t[r_{M,t+1}] - r_{f,t+1} = \lambda_M E_t[\sigma_{M,t+1}^2] + \lambda_L E_t[L_{M,t+1}] + \lambda_{H,x} E_t[\sigma_{M\Delta x,t+1}].$$
(3)

- λ_M implies that investors require compensation for bearing systematic market risk.
- λ_L represents the compensation for bearing aggregate market illiquidity.
- λ_{H,x} accounts for the pricing of unfavorable shifts in the investment opportunity set associated with state variables x.

Intertemporal Pricing Equation

A modified version of the model captures the dynamics of trading and non-trading periods:

$$E_{t}[r_{M,cc,t+1}] - r_{f,t+1} = \lambda_{M,oc} \operatorname{Var}_{t}[r_{M,oc,t+1}] + \lambda_{M,co} \operatorname{Var}_{t}[r_{M,co,t+1}] + \lambda_{M,occo} \operatorname{Cov}_{t}[r_{M,oc,t+1}, r_{M,co,t+1}] + \lambda_{L,oc} E_{t}[L_{M,oc,t+1}] + \lambda_{L,co} E_{t}[L_{M,co,t+1}] + \lambda_{H,x} \operatorname{Cov}_{t}[r_{M,cc,t+1}, \Delta x_{t+1}].$$

$$(4)$$

The expected equity market premium is then a linear function of

- Conditional trading volatility and conditional volatility of jumps induced by trading breaks,
- Conditional covariance between trading and non-trading returns,
- Conditional expectations associated with illiquidity in trading and non-trading periods,
- Conditional covariance associated with innovations in state variables *x* that predict changes in the investment opportunity set.

Empirical Analysis

- TAQ data from January 3, 2007 to December 29, 2023
- Major U.S. ETF: SPDR S&P 500 ETF
- Daily return series:
 - Total return $(r_{M,cc})$: Log-return from official market closure at 4:00 PM to official market closure at 4:00 PM the next day.
 - Trading return ($r_{M,oc}$): Log-return from official market opening at 9:30 AM to official market closure at 4:00 PM.
 - Non-trading return ($r_{M,co}$): Log-return from official market closure at 4:00 PM to the official market opening at 9:30 AM the next day.

- Per-minute bid-ask spreads during three distinct periods each day:
 - Pre-trading hours (*pre*): This period covers the time before the official market opening from 4:00 AM to 9:30 AM.
 - Market-trading hours (*market*): This period represents the official trading hours when the market is open from 9:30 AM to 4.00 PM.
 - Post-trading hours (*post*): This period includes the time after the market officially closes from 4:00 PM to 8:00 PM.

- Macroeconomic variables
 - Default spread (DEF): Difference between the yields on U.S. high-yield corporate bonds (Moody's Baa) and low-yield (Moody's Aaa) corporate bonds.
 - Term spread (TERM): Difference between the yields on the 10-year Treasury bond and 3-month Treasury bill.
- Option-implied volatility: VIX index Chicago Board Options Exchange
- Equity factors
 - Size (SMB) and value (HML) factor (Fama/French 1993)
 - Profitability (RMW) and investment (CMA) factor (Fama/French 2015)

Empirical Results: Return Statistics

	R _{M,cc}	$R_{M,oc}$	R _{M,co}
Min (%)	-12.3638	-8.2173	-11.7082
Mean (%)	0.0290	0.0140	0.0150
Max (%)	10.8385	7.3420	5.8557
Std. Dev. (%)	1.2573	0.9698	0.7591
Skewness	-0.4725	-0.3057	-1.7955
Kurtosis	14.5905	11.3805	30.4936
JB Statistics	24517	12801	139376
	(0.0000)	(0.0000)	(0.0000)
AB Statistic	8752895		
	(0.0000)		
Panel B: Correlations			
	R _{M,cc}	$R_{M,oc}$	R _{M,co}
R _{M,cc}	1		
R _{M,oc}	0.7976	1	
	(0.0000)		
R _{M,co}	0.6373	0.0434	1

Table 1: Descriptive Statistics

Descriptive statistics for daily total market returns ($R_{M,cc}$), trading returns ($R_{M,oc}$) and non-trading returns ($R_{M,co}$) from January 3, 2007 to December 29, 2023 (4278 daily observations). Return calculation is based on the midiquote between bid and ask prices. The Ansari-Bradley (AB) rank sum test of dispersion tests for differences in trading and non-trading return variance. P-values in parentheses.

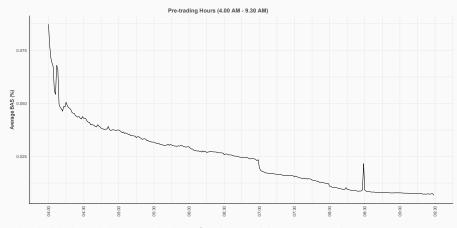
Empirical Results: Intraday Price Pattern



Average intraday midquotes from market opening at 9:30 AM to market closure at 4:00 PM. Midquotes are sampled in 1-minute intervals over the full sample period from January 3, 2007 to December 29, 2023.

Empirical Results: Intraday Illiquidity Pattern

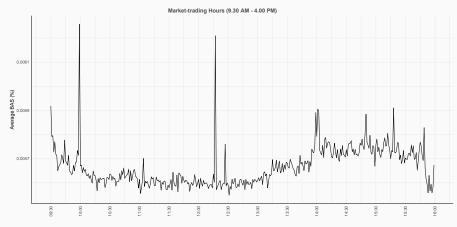
Figure 2: Intraday Illiquidity Pattern (Pre-trading)



Intraday bid-ask spreads during pre-trading hours. Spreads are sampled in 1-minute intervals.

Empirical Results: Intraday Illiquidity Pattern

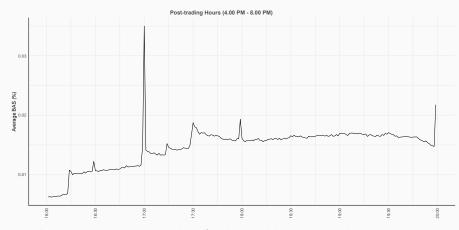
Figure 3: Intraday Illiquidity Pattern (Market-trading)



Intraday bid-ask spreads during market-trading hours. Spreads are sampled in 1-minute intervals.

Empirical Results: Intraday Illiquidity Pattern

Figure 4: Intraday Illiquidity Pattern (Post-trading)



Intraday bid-ask spreads during post-trading hours. Spreads are sampled in 1-minute intervals.

Model I	λ_0	$\lambda_{M,cc}$					Adj. R ²
	0.0003	0.8353					10.53%
	(0.0000)	(0.0001)					
Model II	λ_0	$\lambda_{M,oc}$	$\lambda_{M,co}$	$\lambda_{M,occo}$			Adj.R ²
	0.0003	1.2791	-				9.61%
	(0.0003)	(0.0000)	-				
	0.0003	-	1.8758				9.91%
	(0.0003)	-	(0.0000)				
	0.0003	0.5294	1.9085	-0.0186			10.66%
	(0.0003)	(0.4021)	(0.0000)	(0.0173)			
Model III	λ_0	$\lambda_{M,oc}$	$\lambda_{M,co}$	$\lambda_{M,occo}$	$\lambda_{L,oc}$	$\lambda_{L,\omega}$	Adj. R ²
	0.0009	0.6780	1.8844	-0.0366	0.0078	0.0468	11.39%
	(0.0000)	(0.2178)	(0.0000)	(0.0000)	(0.2462)	(0.0030)	

Table 2: Estimates of Risk Premia in Equation (4)

Modified ICAPM estimation results with $\lambda_{H,x} = 0$. Estimation is done by Maximum-likelihood, p-values based on heteroskedasticity and autocorrelation robust standard errors are in parentheses.

Empirical Results: Stochastic Investment Opportunities

Panel A: Macro	-economic Variabl	es						
λ_0	$\lambda_{M,oc}$	$\lambda_{M,co}$	$\lambda_{M,occo}$	$\lambda_{L,oc}$	$\lambda_{L,co}$	$\lambda_{H,Def}$	$\lambda_{H, Tarm}$	Adj. R ² (%)
0.0001	0.5244	1.7279	-0.0362	0.0074	0.0473	-0.6889	-	11.42
(0.0000)	(0.4170)	(0.0006)	(0.0000)	(0.2549)	(0.0025)	(0.7534)	-	
0.0001	0.6686	1.8984	-0.0367	0.0071	0.0478	-	-0.0186	11.37
(0.0000)	(0.2309)	(0.0000)	(0.0000)	(0.2621)	(0.0053)	-	(0.9216)	
0.0001	0.5222	1.7145	-0.3620	0.0075	0.0469	-0.7118	0.0088	11.40
(0.0000)	(0.4219)	(0.0106)	(0.0000)	(0.2626)	(0.0086)	(0.7730)	(0.9675)	
Panel B: Optio	n-implied Volatility	,						
λ_0	$\lambda_{M,oc}$	$\lambda_{M,co}$	$\lambda_{M,occo}$	$\lambda_{L,oc}$	$\lambda_{L,co}$	$\lambda_{H,VIX}$		Adj. R ² (%)
0.0001	1.19523	2.8231	-0.0468	0.0089	0.0356	0.0558		13.73
(0.0000)	(0.1572)	(0.0000)	(0.0000)	(0.1749)	(0.0228)	(0.0154)		

Table 3: Risk-Return Trade-off under Stochastic Investment Opportunities

Modified ICAPM estimation results with $\lambda_{H,x} \neq 0$. Conditional covariances between market returns and innovations in state variables are estimated by the DCC model of Engle (2002). Robust p-values in parentheses.

λ_0	$\lambda_{M,oc}$	$\lambda_{M,co}$	$\lambda_{M,occo}$	$\lambda_{L,oc}$	$\lambda_{L,co}$	$\lambda_{H,SMB}$	$\lambda_{H,HML}$	$\lambda_{H,RMW}$	$\lambda_{H,CMA}$	Adj.R ² (%)
0.0001	1.1823	2.4231	-0.0468	0.0090	0.0363	0.0580	-	-	-	12.73
(0.0000)	(0.0572)	(0.0000)	(0.0000)	(0.1749)	(0.0228)	(0.0154)	-	-	-	
0.0001	1.3854	2.7709	-0.0379	0.0074	0.0680	-	-0.0234	-	-	13.11
(0.0000)	(0.0095)	(0.0000)	(0.0000)	(0.3254)	(0.0002)	-	(0.0520)	-	-	
0.0001	-0.0509	1.4750	-0.0386	0.0089	0.0368	-	-	-0.4498	-	12.16
(0.0000)	(0.9315)	(0.0003)	(0.0000)	(0.1744)	(0.0165)	-	-	(0.1176)	-	
0.0001	0.3060	1.7333	-0.0406	0.0070	0.0587	-	-	-	-0.0367	11.68
(0.0000)	(0.5942)	(0.0015)	(0.0000)	(0.2725)	(0.0020)	-	-	-	(0.2621)	
0.0001	1.9739	3.4078	-0.0462	0.0086	0.0579	0.0626	-0.0246	-	-	14.67
(0.0000)	(0.0001)	(0.0000)	(0.0000)	(0.2418)	(0.0030)	(0.0093)	(0.0447)	-	-	
0.0001	1.3104	3.1497	-0.0456	0.00871	0.0592	0.0523	-0.0246	-0.0236	-0.0165	14.90
(0.0000)	(0.1727)	(0.0000)	(0.0000)	(0.2128)	(0.0012)	(0.1147)	(0.0479)	(0.5701)	(0.5701)	

Table 3 Cont.: Risk-Return Trade-off under Stochastic Investment Opportunities

Estimation results with $\lambda_{H,x} \neq 0$. Conditional covariances between market returns and innovations in state variables are estimated by the DCC model of Engle (2002). Fama-French (1993, 2015) factors : Small-minus-big (SMB), high-minus-low (HML), robust-minus-weak (RMW), conservative-minus-aggressive (CMA). Robust p-values in parentheses.

Summary & Conclusions

The equity premium compensates primarily for bearing risk and illiquidity during market closures.

 \Rightarrow Conditional volatility and illiquidity during trading hours play only a minor role in shaping the intertemporal risk-return relationship.

Why?

- Expected costs of non-marketability induce a discount prior to the official closure in line with Longstaff's (1995) theory.
- Higher risk aversion of investors during market closures with reduced marketability and increased information asymmetry.
- Intraday patterns suggest a potential 'clientele effect': Assets change owners.

Thank you for your attention!

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