

Internet Appendix to “What Drives Booms and Busts in Value?”

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This Appendix provides a variety of supplemental information for Campbell, Giglio, and Polk (CGP 2023).

Table 1: **Quarterly Vector Autoregression Estimation**

The table shows the weighted least squares (WLS) parameter estimates for a quarterly first-order vector autoregression (VAR) model. The state variables in the VAR are the log real return on the CRSP value-weight index (r_M), the realized variance ($RVAR$) of within-quarter daily simple returns on the CRSP value-weight index, the log ratio of the S&P 500's price to its 10-year moving average of earnings (PE), the term yield spread ($TERM$) in percentage points, measured as the difference between the log yield on 10-year Treasuries and the log yield on 3-month Treasuries, the default yield spread (DEF) in percentage points, measured as the difference between the log yield on Moody's BAA bonds and the log yield on Moody's AAA bonds, and the small-stock value spread (VS), the difference in the log book-to-market ratios of small-value and small-growth stocks. For the sake of interpretation, we estimate the VAR in two stages. Panel A reports the WLS parameter estimates of a first-stage regression forecasting $RVAR$ with the VAR state variables. The forecasted values from this regression are used in the second stage of the estimation procedure as the state variable $EVAR$, replacing $RVAR$ in the second-stage VAR. Panel B reports WLS parameter estimates of the full second-stage VAR. Initial WLS weights on each observation are inversely proportional to $RVAR_t$ and $EVAR_t$ in the first and second stage respectively, and are then shrunk to equal weights so that the maximum ratio of actual weights used is less than or equal to five. In addition, the forecasted values for both $RVAR$ and $EVAR$ are constrained to be positive. In Panels A and B, Column 1–7 report coefficients on an intercept and the six explanatory variables, and Column 8 shows the implied R^2 statistic for the unscaled model. We report t -statistics in parentheses. The sample period for the dependent variables is 1926:3–2022:1, with 384 quarterly data points.

Panel A: Forecasting quarterly realized variance ($RVAR_{t+1}$)							
Constant	$r_{M,t}$	$RVAR_t$	PE_t	$TERM_t$	DEF_t	VS_t	R^2
(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
-0.019	-0.005	0.340	0.005	-0.001	0.006	0.002	33.30%
(-3.48)	(-1.13)	(6.39)	(3.46)	(-1.28)	(4.59)	(0.82)	

Panel B: VAR estimates

Second stage	Constant	$r_{M,t}$	$EVAR_t$	PE_t	$TERM_t$	DEF_t	VS_t	R^2
	(1)	(2)	(3)	(4)	(5)	(6)	(7)	(8)
$r_{M,t+1}$	0.202 (2.69)	0.062 (1.09)	2.332 (1.33)	-0.048 (-2.30)	0.004 (0.76)	-0.019 (-0.84)	-0.027 (-1.33)	2.34%
$EVAR_{t+1}$	-0.016 (-5.31)	-0.002 (-0.98)	0.369 (5.34)	0.004 (5.16)	0.000 (-1.90)	0.005 (5.16)	0.002 (2.47)	56.99%
PE_{t+1}	0.126 (1.66)	0.056 (0.98)	1.418 (0.80)	0.963 (45.44)	0.005 (0.82)	-0.015 (-0.64)	-0.008 (-0.41)	94.18%
$TERM_{t+1}$	-0.014 (-0.04)	-0.040 (-0.14)	4.907 (0.56)	0.019 (0.18)	0.831 (29.76)	0.152 (1.31)	0.001 (0.01)	76.91%
DEF_{t+1}	0.151 (0.85)	-0.355 (-2.65)	3.912 (0.95)	-0.049 (-0.98)	0.001 (0.07)	0.849 (15.55)	0.079 (1.62)	87.46%
VS_{t+1}	0.156 (2.60)	0.070 (1.56)	3.210 (2.30)	-0.018 (-1.09)	-0.004 (-0.95)	-0.005 (-0.29)	0.930 (56.38)	92.57%

Table 2: **HML: 17-Industry Decomposition and CAPM Pricing**

The table reports a decomposition of Fama and French’s (1993) HML into its intra- and inter-industry components. We first create an HML portfolio within each industry in the same way as HML. We then combine these industry-specific HML portfolios using industry market weights. We define industries using Fama and French’s (1997) mapping of SICC into 17 industries, based on historical classifications from Ken French’s website. We scale the resulting composite portfolio so that a full-sample regression of HML on that scaled portfolio, the first regression in the table, has a unit loading. We define the scaled portfolio as HML_{Intra}^{17} and the constant and residual from the regression as HML_{Inter}^{17} . The sample is 1963Q3-2022Q1. We report t -statistics in parentheses.

Panel A: Full-sample estimates						
		constant	HML_{Intra}^{17}	HML	$RMRF$	\widehat{R}^2
(1)	HML	-0.15% (-0.77)	1.00 (27.26)			76.0%
(2)	HML_{Intra}^{17}	0.37% (2.18)		0.76 (27.26)		76.0%
(3)	HML_{Inter}^{17}	-0.37% (-2.18)		0.24 (8.55)		23.6%
(4)	HML	1.24% (3.16)			-0.17 (-3.83)	5.5%
(5)	HML_{Intra}^{17}	1.36% (3.99)			-0.16 (-3.98)	6.0%
(6)	HML_{Inter}^{17}	-0.12% (-0.62)			-0.02 (-0.72)	-0.2%

Panel B: Pre-1990 estimates

		constant	HML _{Intra} ¹⁷	HML	<i>RMRF</i>	\widehat{R}^2
(1)	HML	-0.10% (-0.37)	0.98 (17.75)			74.8%
(2)	HML _{Intra} ¹⁷	0.47% (2.02)		0.77 (17.75)		74.8%
(3)	HML _{Inter} ¹⁷	-0.47% (-2.02)		0.23 (5.42)		21.1%
(4)	HML	1.75% (3.76)			-0.25 (-4.66)	16.4%
(5)	HML _{Intra} ¹⁷	1.83% (4.35)			-0.20 (-4.17)	13.4%
(6)	HML _{Inter} ¹⁷	-0.07% (-0.29)			-0.05 (-1.68)	1.7%

Panel C: Post-1990 estimates

		constant	HML _{Intra} ¹⁷	HML	<i>RMRF</i>	\widehat{R}^2
(1)	HML	-0.18% (-0.61)	1.01 (20.25)			76.3%
(2)	HML _{Intra} ¹⁷	0.29% (1.16)		0.76 (20.25)		76.3%
(3)	HML _{Inter} ¹⁷	-0.29% (-1.16)		0.24 (6.50)		24.5%
(4)	HML	0.72% (1.18)			-0.10 (-1.42)	0.8%
(5)	HML _{Intra} ¹⁷	0.92% (1.75)			-0.11 (-1.86)	1.9%
(6)	HML _{Inter} ¹⁷	-0.20% (-0.67)			0.01 (0.36)	-0.7%

Table 3: **HML Alternative Decomposition 1 and CAPM Pricing**

The table reports a decomposition of Fama and French’s (1993) HML into its intra- and inter-industry components using our first alternative to the baseline approach studied in the main text. Rather than create a within-industry *HML*, we create an industry-demeaned *HML*. We create a value factor in the same way as *HML*, but we first industry-adjust firm-level BE/ME using Fama and French’s (1997) mapping of SICC into 48 industries, based on historical classifications from Ken French’s website. As this process still results in some incidental industry exposure, we then industry-neutralize this portfolio. In particular, we compute its industry exposure and using the resulting portfolio to offset any remaining industry exposure by differencing. We scale the resulting industry-neutralized, industry-demeaned portfolio so that a regression of *HML* on that portfolio, the first regression in the table, has a unit loading. We define the scaled portfolio as HML_{Intra}^{Alt1} and the constant and residual from the regression as HML_{Inter}^{Alt1} . The sample is 1963Q3-2022Q1. We report *t*-statistics in parentheses.

Panel A: Full-sample estimates						
		constant	HML_{Intra}^{Alt1}	HML	<i>RMRF</i>	\widehat{R}^2
(1)	HML	-0.17% (-0.78)	1.00 (23.58)			70.3%
(2)	HML_{Intra}^{Alt1}	0.45% (2.43)		0.70 (23.58)		70.3%
(3)	HML_{Inter}^{Alt1}	-0.45% (-2.43)		0.30 (9.88)		29.2%
(4)	HML	1.24% (3.16)			-0.17 (-3.83)	5.5%
(5)	HML_{Intra}^{Alt1}	1.15% (3.39)			-0.03 (-0.76)	-0.2%
(6)	HML_{Inter}^{Alt1}	0.09% (0.43)			-0.14 (-6.11)	13.4%

Panel B: Pre-1990 estimates

		constant	HML _{Intra} ^{Alt1}	HML	<i>RMRF</i>	\widehat{R}^2
(1)	HML	0.02% (0.08)	0.95 (17.09)			73.3%
(2)	HML _{Intra} ^{Alt1}	0.38% (1.56)		0.77 (17.09)		73.3%
(3)	Inter HML _{Inter} ^{Alt1}	-0.38% (-1.56)		0.23 (5.00)		18.5%
(4)	HML	1.75% (3.76)			-0.25 (-4.66)	16.4%
(5)	HML _{Intra} ^{Alt1}	1.71% (3.90)			-0.17 (-3.40)	9.0%
(6)	HML _{Inter} ^{Alt1}	0.04% (0.17)			-0.08 (-2.68)	5.5%

Panel C: Post-1990 estimates

		constant	HML _{Intra} ^{Alt1}	HML	<i>RMRF</i>	\widehat{R}^2
(1)	HML	-0.29% (-0.88)	1.03 (16.68)			68.6%
(2)	HML _{Intra} ^{Alt1}	0.43% (1.62)		0.67 (16.68)		68.6%
(3)	HML _{Inter} ^{Alt1}	-0.43% (-1.62)		0.33 (8.25)		34.6%
(4)	HML	0.72% (1.18)			-0.10 (-1.42)	0.8%
(5)	HML _{Intra} ^{Alt1}	0.53% (1.07)			0.10 (1.80)	1.7%
(6)	HML _{Inter} ^{Alt1}	0.20% (0.64)			-0.20 (-5.70)	19.9%

Table 4: **HML Alternative Decomposition 2 and CAPM Pricing**

The table reports a decomposition of Fama and French’s (1993) *HML* into its intra- and inter-industry components using our second alternative to the baseline approach in the main text. Rather than create a within-industry *HML*, we industry neutralize *HML*, namely, computing the industry exposure of *HML* and using that portfolio to offset *HML*’s industry exposure by differencing. We measure industry using Fama and French’s (1997) mapping of SICC into 48 industries, based on historical classifications from Ken French’s website. We scale the resulting industry-neutralized portfolio so that a regression of *HML* on that portfolio, the first regression in the table, has a unit loading. We define the scaled portfolio as $HML_{\text{Intra}}^{\text{Alt2}}$ and the constant and residual from the regression as $HML_{\text{Inter}}^{\text{Alt2}}$. The sample is 1963Q3-2022Q1. We report *t*-statistics in parentheses.

Panel A: Full-sample estimates						
		constant	$HML_{\text{Intra}}^{\text{Alt2}}$	<i>HML</i>	<i>RMRF</i>	\widehat{R}^2
(1)	<i>HML</i>	-0.32% (-1.57)	1.00 (26.94)			75.6%
(2)	$HML_{\text{Intra}}^{\text{Alt2}}$	0.54% (3.15)		0.76 (26.94)		75.6%
(3)	$HML_{\text{Inter}}^{\text{Alt2}}$	-0.54% (-3.15)		0.24 (8.65)		24.0%
(4)	<i>HML</i>	1.24% (3.16)			-0.17 (-3.83)	5.5%
(5)	$HML_{\text{Intra}}^{\text{Alt2}}$	1.58% (4.71)			-0.19 (-4.83)	8.7%
(6)	$HML_{\text{Inter}}^{\text{Alt2}}$	-0.34% (-1.71)			0.01 (0.60)	-0.3%

Panel B: Pre-1990 estimates

		constant	HML _{Intra} ^{Alt2}	HML	<i>RMRF</i>	\widehat{R}^2
(1)	HML	-0.19% (-0.74)	0.94 (18.45)			76.2%
(2)	HML _{Intra} ^{Alt2}	0.57% (2.40)		0.81 (18.45)		76.2%
(3)	HML _{Inter} ^{Alt2}	-0.57% (-2.40)		0.19 (4.35)		14.5%
(4)	HML	1.75% (3.76)			-0.25 (-4.66)	16.4%
(5)	HML _{Intra} ^{Alt2}	2.07% (5.00)			-0.27 (-5.71)	23.0%
(6)	HML _{Inter} ^{Alt2}	-0.32% (-1.28)			0.02 (0.76)	-0.4%

Panel C: Post-1990 estimates

		constant	HML _{Intra} ^{Alt2}	HML	<i>RMRF</i>	\widehat{R}^2
(1)	HML	-0.36% (-1.23)	1.03 (19.61)			75.1%
(2)	HML _{Intra} ^{Alt2}	0.47% (1.89)		0.73 (19.61)		75.1%
(3)	HML _{Inter} ^{Alt2}	-0.47% (-1.89)		0.27 (7.31)		29.2%
(4)	HML	0.72% (1.18)			-0.10 (-1.42)	0.8%
(5)	HML _{Intra} ^{Alt2}	1.07% (2.10)			-0.11 (-1.82)	1.8%
(6)	HML _{Inter} ^{Alt2}	-0.35% (-1.15)			0.01 (0.20)	-0.8%

Table 5: **Cash-flow, Discount-rate, and Variance Betas Using Log Returns**

The table shows the estimated market ($\hat{\beta}$), cash-flow ($\hat{\beta}_{CF}$), discount-rate ($\hat{\beta}_{DR}$), and variance ($\hat{\beta}_V$) betas for HML and its intra- and inter-industry components when measured using log, rather than simple, returns. The left side of the table runs simple regressions while the right side of the table estimates a multiple regression with all three ICAPM News terms as regressors. The resulting point estimates in both the simple and multiple ICAPM regressions are scaled as in Campbell, Giglio, Polk, and Turley (2018). The sample is 1963Q3-2022Q1, which is then split into two sub-samples in Panels B and C. We report t -statistics in parentheses.

		simple regressions				multiple regression			
		$\hat{\beta}$	$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	\widehat{R}^2
Panel A: Full Sample									
(1)	HML	-0.15 (-3.35)	-0.23 (-6.56)	0.08 (6.20)	-0.11 (-10.24)	-0.26 (-7.80)	0.09 (7.46)	-0.06 (-5.00)	48.1%
(2)	HML _{Intra}	-0.06 (-1.69)	-0.14 (-4.34)	0.07 (6.25)	-0.10 (-10.08)	-0.14 (-4.47)	0.07 (5.72)	-0.06 (-5.76)	39.2%
(3)	HML _{Inter}	-0.08 (-3.54)	-0.10 (-4.89)	0.01 (1.60)	-0.02 (-2.62)	-0.12 (-5.34)	0.03 (3.22)	0.00 (0.41)	12.7%
Panel B: Pre-1990									
(4)	HML	-0.25 (-4.70)	-0.28 (-6.96)	0.04 (2.16)	-0.10 (-7.96)	-0.27 (-4.82)	0.07 (4.49)	-0.03 (-2.15)	49.7%
(5)	HML _{Intra}	-0.16 (-3.23)	-0.20 (-4.90)	0.04 (2.33)	-0.07 (-5.99)	-0.19 (-3.24)	0.06 (3.67)	-0.03 (-1.66)	33.5%
(6)	HML _{Inter}	-0.09 (-2.44)	-0.09 (-2.92)	0.00 (0.09)	-0.02 (-2.69)	-0.08 (-1.66)	0.01 (0.78)	-0.01 (-0.55)	6.3%
Panel C: Post-1990									
(7)	HML	-0.05 (-0.75)	-0.19 (-3.27)	0.14 (6.68)	-0.13 (-6.91)	-0.27 (-5.91)	0.13 (6.22)	-0.06 (-3.29)	48.4%
(8)	HML _{Intra}	0.03 (0.61)	-0.08 (-1.62)	0.11 (6.87)	-0.12 (-7.90)	-0.13 (-3.31)	0.09 (4.87)	-0.07 (-4.47)	43.6%
(9)	HML _{Inter}	-0.09 (-2.66)	-0.11 (-4.05)	0.02 (2.02)	-0.01 (-1.28)	-0.14 (-5.08)	0.04 (3.39)	0.01 (0.85)	18.0%

Table 6: **Cash-flow, Discount-rate, and Variance Betas: 17 industries**

The table shows the estimated market ($\widehat{\beta}$), cash-flow ($\widehat{\beta}_{CF}$), discount-rate ($\widehat{\beta}_{DR}$), and variance ($\widehat{\beta}_V$) betas for HML and its intra- and inter-industry components (based on Fama and French's 17 industries). The left side of the table runs simple regressions while the right side of the table estimates a multiple regression with all three ICAPM News terms as regressors. The resulting point estimates in both the simple and multiple ICAPM regressions are scaled as in Campbell, Giglio, Polk, and Turley (2018). The sample is 1963Q3-2022Q1, which is then split into two sub-samples in Panels B (1963Q3-1990Q1) and C (1990Q2-2022Q1). We report t -statistics in parentheses.

		simple regressions				multiple regression			
		$\widehat{\beta}$	$\widehat{\beta}_{DR}$	$\widehat{\beta}_{CF}$	$\widehat{\beta}_V$	$\widehat{\beta}_{DR}$	$\widehat{\beta}_{CF}$	$\widehat{\beta}_V$	\widehat{R}^2
Panel A: Full Sample									
(1)	HML	-0.16 (-3.51)	-0.24 (-6.77)	0.09 (6.14)	-0.11 (-10.07)	-0.27 (-8.13)	0.10 (7.61)	-0.06 (-4.77)	48.40%
(2)	HML ¹⁷ _{Intra}	-0.15 (-3.91)	-0.21 (-6.76)	0.06 (4.76)	-0.09 (-9.07)	-0.23 (-7.17)	0.07 (5.73)	-0.05 (-4.33)	40.31%
(3)	HML ¹⁷ _{Inter}	-0.01 (-0.23)	-0.03 (-1.63)	0.03 (3.65)	-0.02 (-3.36)	-0.05 (-2.16)	0.03 (3.42)	-0.01 (-1.07)	8.21%
Panel B: Pre-1990									
(4)	HML	-0.25 (-4.75)	-0.29 (-7.04)	0.04 (2.15)	-0.10 (-8.00)	-0.28 (-4.91)	0.07 (4.54)	-0.04 (-2.13)	50.23%
(5)	HML ¹⁷ _{Intra}	-0.22 (-4.64)	-0.25 (-6.64)	0.03 (1.77)	-0.08 (-6.76)	-0.26 (-4.78)	0.06 (3.99)	-0.02 (-1.27)	42.66%
(6)	HML ¹⁷ _{Inter}	-0.03 (-1.16)	-0.04 (-1.78)	0.01 (1.13)	-0.02 (-2.72)	-0.02 (-0.53)	0.01 (1.08)	-0.02 (-1.34)	4.95%
Panel C: Post-1990									
(7)	HML	-0.06 (-0.92)	-0.20 (-3.46)	0.14 (6.56)	-0.13 (-6.71)	-0.29 (-6.18)	0.13 (6.30)	-0.06 (-3.09)	48.49%
(8)	HML ¹⁷ _{Intra}	-0.09 (-1.40)	-0.18 (-3.60)	0.09 (4.99)	-0.11 (-6.23)	-0.23 (-5.33)	0.09 (4.34)	-0.06 (-3.30)	39.09%
(9)	HML ¹⁷ _{Inter}	0.02 (0.60)	-0.02 (-0.71)	0.04 (3.75)	-0.02 (-2.20)	-0.05 (-1.81)	0.05 (3.47)	0.00 (0.02)	10.33%

Table 7: **ICAPM Betas of HML Alternative Decomposition 1**

The table shows the estimated market ($\hat{\beta}$), cash-flow ($\hat{\beta}_{CF}$), discount-rate ($\hat{\beta}_{DR}$), and variance ($\hat{\beta}_V$) betas for HML and its intra- and inter-industry components using our first alternative to the baseline approach studied in the main text. The left side of the table runs simple regressions while the right side of the table estimates a multiple regression with all three ICAPM News terms as regressors. The resulting point estimates in both the simple and multiple ICAPM regressions are scaled as in Campbell, Giglio, Polk, and Turley (2018). The sample is 1963Q3-2022Q1, which is then split into two sub-samples in Panels B and C. We report t -statistics in parentheses.

		simple regressions				multiple regression			
		$\hat{\beta}$	$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	\widehat{R}^2
Panel A: Full Sample									
(1)	HML	-0.16 (-3.51)	-0.24 (-6.77)	0.09 (6.14)	-0.11 (-10.07)	-0.27 (-8.13)	0.10 (7.61)	-0.06 (-4.77)	48.4%
(2)	HML _{Intra} ^{Alt1}	-0.04 (-1.04)	-0.12 (-3.86)	0.08 (7.39)	-0.10 (-10.64)	-0.13 (-4.36)	0.08 (6.76)	-0.06 (-6.14)	43.5%
(3)	HML _{Inter} ^{Alt1}	-0.12 (-4.94)	-0.12 (-6.02)	0.00 (0.20)	-0.01 (-2.01)	-0.14 (-6.14)	0.02 (2.38)	0.01 (0.90)	14.5%
Panel B: Pre-1990									
(4)	HML	-0.25 (-4.75)	-0.29 (-7.04)	0.04 (2.15)	-0.10 (-8.00)	-0.28 (-4.91)	0.07 (4.54)	-0.04 (-2.13)	50.2%
(5)	HML _{Intra} ^{Alt1}	-0.20 (-3.95)	-0.23 (-5.91)	0.04 (2.30)	-0.08 (-6.60)	-0.24 (-4.32)	0.06 (4.29)	-0.02 (-1.43)	41.1%
(6)	HML _{Inter} ^{Alt1}	-0.06 (-1.95)	-0.06 (-2.36)	0.00 (0.15)	-0.02 (-2.60)	-0.04 (-0.94)	0.00 (0.45)	-0.01 (-1.01)	4.1%
Panel C: Post-1990									
(7)	HML	-0.06 (-0.92)	-0.20 (-3.46)	0.14 (6.56)	-0.13 (-6.71)	-0.29 (-6.18)	0.13 (6.30)	-0.06 (-3.09)	48.5%
(8)	HML _{Intra} ^{Alt1}	0.11 (1.91)	-0.02 (-0.50)	0.13 (8.69)	-0.12 (-8.16)	-0.09 (-2.40)	0.11 (6.17)	-0.07 (-4.33)	48.7%
(9)	HML _{Inter} ^{Alt1}	-0.17 (-4.73)	-0.18 (-5.86)	0.00 (0.23)	-0.01 (-0.79)	-0.20 (-6.18)	0.03 (1.95)	0.01 (0.57)	22.1%

Table 8: **ICAPM Betas of HML Alternative Decomposition 2**

The table shows the estimated market ($\hat{\beta}$), cash-flow ($\hat{\beta}_{CF}$), discount-rate ($\hat{\beta}_{DR}$), and variance ($\hat{\beta}_V$) betas for HML and its intra- and inter-industry components using our second alternative to the baseline approach studied in the main text. The left side of the table runs simple regressions while the right side of the table estimates a multiple regression with all three ICAPM News terms as regressors. The resulting point estimates in both the simple and multiple ICAPM regressions are scaled as in Campbell, Giglio, Polk, and Turley (2018). The sample is 1963Q3-2022Q1, which is then split into two sub-samples in Panels B and C. We report t -statistics in parentheses.

		simple regressions				multiple regression			
		$\hat{\beta}$	$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	\widehat{R}^2
Panel A: Full Sample									
(1)	HML	-0.16 (-3.51)	-0.24 (-6.77)	0.09 (6.14)	-0.11 (-10.07)	-0.27 (-8.13)	0.10 (7.61)	-0.06 (-4.77)	48.4%
(2)	HML _{Intra} ^{Alt2}	-0.18 (-4.78)	-0.23 (-7.51)	0.05 (3.73)	-0.11 (-10.99)	-0.21 (-6.80)	0.05 (4.03)	-0.07 (-6.59)	44.6%
(3)	HML _{Inter} ^{Alt2}	0.03 (1.11)	-0.01 (-0.66)	0.04 (5.48)	-0.01 (-1.31)	-0.07 (-3.14)	0.05 (6.24)	0.01 (1.85)	14.0%
Panel B: Pre-1990									
(4)	HML	-0.25 (-4.75)	-0.29 (-7.04)	0.04 (2.15)	-0.10 (-8.00)	-0.28 (-4.91)	0.07 (4.54)	-0.04 (-2.13)	50.2%
(5)	HML _{Intra} ^{Alt2}	-0.28 (-5.99)	-0.29 (-8.01)	0.01 (0.72)	-0.09 (-8.04)	-0.25 (-4.67)	0.04 (2.73)	-0.04 (-2.27)	47.5%
(6)	HML _{Inter} ^{Alt2}	0.03 (0.97)	0.00 (0.08)	0.03 (3.07)	-0.01 (-0.92)	-0.03 (-0.69)	0.03 (2.91)	0.00 (0.04)	6.5%
Panel C: Post-1990									
(7)	HML	-0.06 (-0.92)	-0.20 (-3.46)	0.14 (6.56)	-0.13 (-6.71)	-0.29 (-6.18)	0.13 (6.30)	-0.06 (-3.09)	48.5%
(8)	HML _{Intra} ^{Alt2}	-0.09 (-1.51)	-0.17 (-3.60)	0.08 (4.57)	-0.12 (-7.62)	-0.20 (-5.00)	0.06 (3.19)	-0.09 (-5.05)	42.6%
(9)	HML _{Inter} ^{Alt2}	0.02 (0.70)	-0.03 (-0.86)	0.05 (4.58)	-0.01 (-0.94)	-0.08 (-2.80)	0.07 (5.56)	0.03 (2.22)	19.0%

Table 9: **HML and Quarterly/Daily ICAPM News: Log Returns**

The table shows the estimated market ($\hat{\beta}$), cash-flow ($\hat{\beta}_{CF}$), discount-rate ($\hat{\beta}_{DR}$), and variance ($\hat{\beta}_V$) betas for HML (Panel A), HML_{Intra} (Panel B), and HML_{Inter} (Panel C) from a multiple regression with all three ICAPM News terms as regressors and using log, rather than simple, returns. In each Panel, the first regression uses quarterly news terms from the quarterly VAR; the second regression uses quarterly news terms constructed by summing daily news terms from the daily VAR; and the third regression uses daily news terms from the daily VAR. We scale the resulting point estimates as in Campbell, Giglio, Polk, and Turley (2018). The sample is 1990Q2-2022Q1. We report t -statistics in parentheses.

Frequency			Multiple Regression			
Regression	VAR		$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	\widehat{R}^2
Panel A: HML						
(1)	Quarterly	Quarterly	-0.27 (-5.91)	0.13 (6.22)	-0.06 (-3.29)	48.42%
(2)	Quarterly	Daily	-0.19 (-4.43)	0.10 (6.23)	-0.02 (-1.62)	41.25%
(3)	Daily	Daily	-0.07 (-1.47)	0.09 (6.80)	-0.01 (-0.69)	27.83%
Panel B: HML _{Intra}						
(4)	Quarterly	Quarterly	-0.13 (-3.31)	0.09 (4.87)	-0.07 (-4.47)	43.64%
(5)	Quarterly	Daily	-0.07 (-1.88)	0.07 (5.01)	-0.04 (-2.97)	38.54%
(6)	Daily	Daily	-0.01 (-0.18)	0.06 (6.83)	-0.02 (-1.90)	24.32%
Panel C: HML _{Inter}						
(7)	Quarterly	Quarterly	-0.14 (-5.08)	0.04 (3.39)	0.01 (0.85)	17.96%
(8)	Quarterly	Daily	-0.13 (-5.01)	0.03 (3.54)	0.01 (1.46)	17.28%
(9)	Daily	Daily	-0.07 (-1.94)	0.03 (3.27)	0.01 (0.71)	14.60%

Table 10: **Explaining HML with Quarterly/Daily ICAPM News: 17 Inds.**

The table shows the estimated market ($\hat{\beta}$), cash-flow ($\hat{\beta}_{CF}$), discount-rate ($\hat{\beta}_{DR}$), and variance ($\hat{\beta}_V$) betas for HML (Panel A), HML_{Intra}^{17} (Panel B), and HML_{Inter}^{17} (Panel C) from a multiple regression with all three ICAPM News terms as regressors. The industry decomposition is based on Fama and French's 17 industries. In each Panel, the first regression uses quarterly news terms from the quarterly VAR; the second regression uses quarterly news terms constructed by summing daily news terms from the daily VAR; and the third regression uses daily news terms from the daily VAR. We scale the resulting point estimates as in Campbell, Giglio, Polk, and Turley (2018). The daily regressions estimating ICAPM betas include 59 lags (i.e., roughly a quarter of daily news); we report the sum of the resulting 60 coefficients associated with each news term. The sample is 1990Q2-2022Q1. We report t -statistics in parentheses.

Frequency			Multiple Regression			
Regression	VAR		$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	\widehat{R}^2
Panel A: HML						
(1)	Quarterly	Quarterly	-0.29 (-6.18)	0.13 (6.30)	-0.06 (-3.09)	48.49%
(2)	Quarterly	Daily	-0.21 (-4.71)	0.10 (6.23)	-0.02 (-1.58)	41.54%
(3)	Daily	Daily	-0.08 (-1.63)	0.09 (6.68)	-0.02 (-0.74)	27.78%
Panel B: HML_{Intra}^{17}						
(4)	Quarterly	Quarterly	-0.23 (-5.33)	0.09 (4.34)	-0.06 (-3.30)	39.09%
(5)	Quarterly	Daily	-0.17 (-4.17)	0.07 (4.50)	-0.03 (-2.11)	33.43%
(6)	Daily	Daily	-0.03 (-0.93)	0.06 (5.98)	-0.03 (-1.79)	26.06%
Panel C: HML_{Inter}^{17}						
(7)	Quarterly	Quarterly	-0.05 (-1.81)	0.05 (3.47)	0.00 (0.02)	10.33%
(8)	Quarterly	Daily	-0.04 (-1.42)	0.03 (3.42)	0.01 (0.61)	8.75%
(9)	Daily	Daily	-0.05 (-1.41)	0.03 (3.44)	0.01 (0.82)	17.92%

Table 11: **HML and Quarterly/Daily ICAPM News: Alt. Decomposition 1**

The table shows the estimated market ($\hat{\beta}$), cash-flow ($\hat{\beta}_{CF}$), discount-rate ($\hat{\beta}_{DR}$), and variance ($\hat{\beta}_V$) betas for HML (Panel A), HML^{Alt1}_{Intra} (Panel B), and HML^{Alt1}_{Inter} (Panel C) from a multiple regression with all three ICAPM News terms as regressors and using our first alternative decomposition of HML. In each Panel, the first regression uses quarterly news terms from the quarterly VAR; the second regression uses quarterly news terms constructed by summing daily news terms from the daily VAR; and the third regression uses daily news terms from the daily VAR. We scale the resulting point estimates as in Campbell, Giglio, Polk, and Turley (2018). The sample is 1990Q2-2022Q1. We report t -statistics in parentheses.

Frequency			Multiple Regression			
Regression	VAR		$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	\widehat{R}^2
Panel A: HML						
(1)	Quarterly	Quarterly	-0.29 (-6.18)	0.13 (6.30)	-0.06 (-3.09)	48.5%
(2)	Quarterly	Daily	-0.21 (-4.71)	0.10 (6.23)	-0.02 (-1.58)	41.5%
(3)	Daily	Daily	-0.08 (-1.63)	0.09 (6.68)	-0.02 (-0.74)	27.8%
Panel B: HML ^{Alt1} _{Intra}						
(4)	Quarterly	Quarterly	-0.09 (-2.45)	0.11 (6.14)	-0.07 (-4.29)	48.4%
(5)	Quarterly	Daily	-0.03 (-0.80)	0.08 (5.68)	-0.03 (-2.58)	40.9%
(6)	Daily	Daily	0.07 (2.39)	0.05 (5.80)	-0.02 (-1.83)	24.6%
Panel C: HML ^{Alt1} _{Inter}						
(7)	Quarterly	Quarterly	-0.19 (-6.12)	0.03 (1.94)	0.01 (0.55)	21.8%
(8)	Quarterly	Daily	-0.18 (-6.38)	0.03 (2.53)	0.01 (0.82)	23.7%
(9)	Daily	Daily	-0.16 (-4.81)	0.03 (3.37)	0.01 (0.87)	11.7%

Table 12: **HML and Quarterly/Daily ICAPM News: Alt. Decomposition 2**

The table shows the estimated market ($\hat{\beta}$), cash-flow ($\hat{\beta}_{CF}$), discount-rate ($\hat{\beta}_{DR}$), and variance ($\hat{\beta}_V$) betas for HML (Panel A), HML^{Alt2}_{Intra} (Panel B), and HML^{Alt2}_{Inter} (Panel C) from a multiple regression with all three ICAPM News terms as regressors and using our second alternative decomposition of HML. In each Panel, the first regression uses quarterly news terms from the quarterly VAR; the second regression uses quarterly news terms constructed by summing daily news terms from the daily VAR; and the third regression uses daily news terms from the daily VAR. We scale the resulting point estimates as in Campbell, Giglio, Polk, and Turley (2018). The sample is 1990Q2-2022Q1. We report t -statistics in parentheses.

Frequency			Multiple Regression			
Regression	VAR		$\hat{\beta}_{DR}$	$\hat{\beta}_{CF}$	$\hat{\beta}_V$	\widehat{R}^2
Panel A: HML						
(1)	Quarterly	Quarterly	-0.29 (-6.18)	0.13 (6.30)	-0.06 (-3.09)	48.5%
(2)	Quarterly	Daily	-0.21 (-4.71)	0.10 (6.23)	-0.02 (-1.58)	41.5%
(3)	Daily	Daily	-0.08 (-1.63)	0.09 (6.68)	-0.02 (-0.74)	27.8%
Panel B: HML ^{Alt2} _{Intra}						
(4)	Quarterly	Quarterly	-0.21 (-4.98)	0.06 (3.19)	-0.09 (-5.04)	42.5%
(5)	Quarterly	Daily	-0.16 (-3.90)	0.06 (4.27)	-0.03 (-2.34)	32.8%
(6)	Daily	Daily	-0.02 (-0.81)	0.04 (5.90)	-0.03 (-2.79)	24.4%
Panel C: HML ^{Alt2} _{Inter}						
(7)	Quarterly	Quarterly	-0.08 (-2.75)	0.07 (5.51)	0.03 (2.29)	18.6%
(8)	Quarterly	Daily	-0.09 (-32.55)	0.03 (33.18)	-0.01 (-14.58)	22.4%
(9)	Daily	Daily	-0.05 (-1.40)	0.03 (3.37)	0.02 (1.59)	11.4%

Table 13: **Daily Vector Autoregression Estimation**

The table shows the ordinary least squares (OLS) parameter estimates for a daily first-order vector autoregression (VAR) model. The state variables in the VAR are the log real return on the CRSP value-weight index (r_M), the squared simple nominal return ($RVAR$) on the CRSP value-weight index, the log ratio of the S&P 500's price to its 10-year moving average of earnings (PE), the term yield spread ($TERM$) in percentage points, measured as the difference between the log yield on 10-year Treasuries and the log yield on 3-month Treasuries, the default yield spread (DEF) in percentage points, measured as the difference between the log yield on Moody's BAA bonds and the log yield on Moody's AAA bonds, the small-stock value spread (VS), the difference in the log book-to-market ratios of small-value and small-growth stocks, the log real return accumulated over the last 60 trading days (r_M^{60}), the realized variance ($RVAR$) of daily simple returns on the CRSP value-weight index over the last 60 trading days, and the squared value of the VIX. Columns 1–10 report coefficients on an intercept and the nine explanatory variables, and Column 11 shows the R^2 statistic. We report t -statistics in parentheses. The sample period for the dependent variables is January 4, 1990–March 31, 2022, with 8,124 daily data points.

	Constant (1)	$r_{M,t}$ (2)	R_t^2 (3)	PE_t (4)	$TERM_t$ (5)	DEF_t (6)	VS_t (7)	$r_{M,t}^{60}$ (8)	$RVAR_t$ (9)	VIX_t^2 (10)	R^2 (11)
$r_{M,t+1}$	0.0087 (3.05)	-0.0428 (-1.40)	0.8360 (1.28)	-0.0023 (-2.68)	-0.0002 (-1.68)	-0.0015 (-1.67)	0.0003 (0.38)	0.0013 (0.55)	-0.0046 (-0.15)	0.0103 (0.81)	0.52%
R_{t+1}^2	-0.0001 (-0.57)	-0.0005 (-0.43)	0.0009 (0.01)	0.0000 (-0.85)	0.0000 (-1.08)	0.0000 (-0.12)	0.0001 (1.63)	-0.0003 (-2.78)	-0.0057 (-2.66)	0.0058 (5.67)	28.46%
PE_{t+1}	0.0087 (3.02)	-0.0733 (-2.43)	0.8830 (1.33)	0.9976 (1159.41)	-0.0002 (-1.37)	-0.0017 (-1.76)	0.0006 (0.68)	0.0012 (0.48)	-0.0144 (-0.47)	0.0124 (0.96)	99.76%
$TERM_{t+1}$	0.0193 (1.36)	-0.2350 (-2.43)	8.2898 (2.94)	-0.0073 (-1.95)	0.9972 (1399.59)	0.0036 (0.96)	0.0038 (1.22)	0.0180 (1.49)	-0.2913 (-2.25)	0.0452 (0.91)	99.68%
DEF_{t+1}	0.0147 (2.48)	-0.0319 (-0.93)	-1.6620 (-1.12)	-0.0039 (-2.35)	-0.0003 (-1.04)	0.9923 (505.43)	0.0021 (1.53)	-0.0197 (-3.66)	-0.1780 (-2.95)	0.1032 (4.37)	99.67%
VS_{t+1}	.0075 (2.68)	0.0368 (3.18)	-0.5394 (-1.07)	-0.0013 (-1.66)	-0.0003 (-3.19)	-0.0007 (-1.02)	0.9984 (1185.32)	0.0029 (1.30)	0.0098 (0.39)	0.0133 (1.90)	99.75%
$r_{M,t+1}^{60}$	0.0091 (2.34)	-0.0246 (-0.69)	1.0073 (1.16)	-0.0020 (-1.74)	-0.0002 (-1.32)	-0.0013 (-1.05)	0.0000 (-0.05)	0.9779 (260.94)	0.1084 (2.03)	-0.0302 (-1.96)	95.86%
$RVAR_{t+1}$	-0.0002 (-0.70)	0.0013 (1.09)	0.0230 (0.22)	0.0000 (0.18)	0.0000 (0.16)	-0.0001 (-0.77)	0.0001 (1.05)	-0.0017 (-4.37)	0.9791 (182.94)	0.0059 (4.85)	99.81%
VIX_{t+1}^2	0.0086 (-3.69)	0.1111 (1.92)	0.1276 (0.09)	0.0025 (3.22)	0.0001 (1.05)	0.0028 (2.94)	-0.0002 (-0.16)	-0.0037 (-1.29)	0.0585 (1.38)	0.9420 (50.43)	93.23%

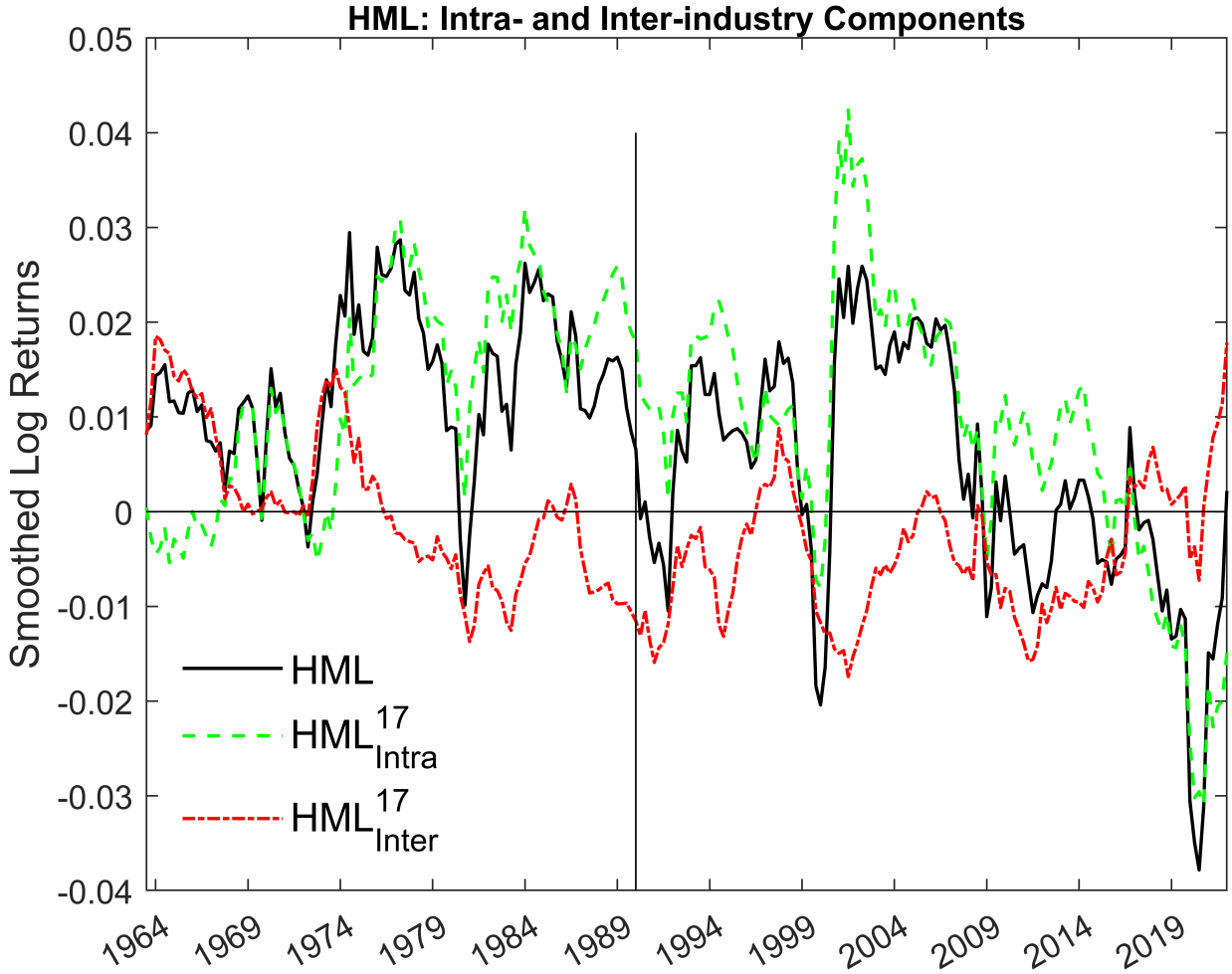


Figure 1: We plot the results from the Appendix Table 2 decomposition of HML into its intra- and inter-industry components for the Compustat period from 1963Q3-2022Q1 based on Fama and French’s classification of firms into 17 industries. The solid black line shows the smoothed log return to Fama and French’s (1993) HML; the dashed green line shows the smoothed log return to the intra-industry component of HML; and the dashed-dotted red line shows the smoothed log return to the inter-industry component. The series are smoothed with a trailing exponentially weighted moving average in which the decay parameter is set to 0.08 per quarter, and the smoothed series is generated, for example, as $MA_t(\text{HML}) = 0.08\text{HML}_t + (1-0.08)MA_{t-1}(\text{HML})$. This decay parameter implies a half-life of two years. The vertical line indicates the start of the subperiod 1990Q2-2022Q1.

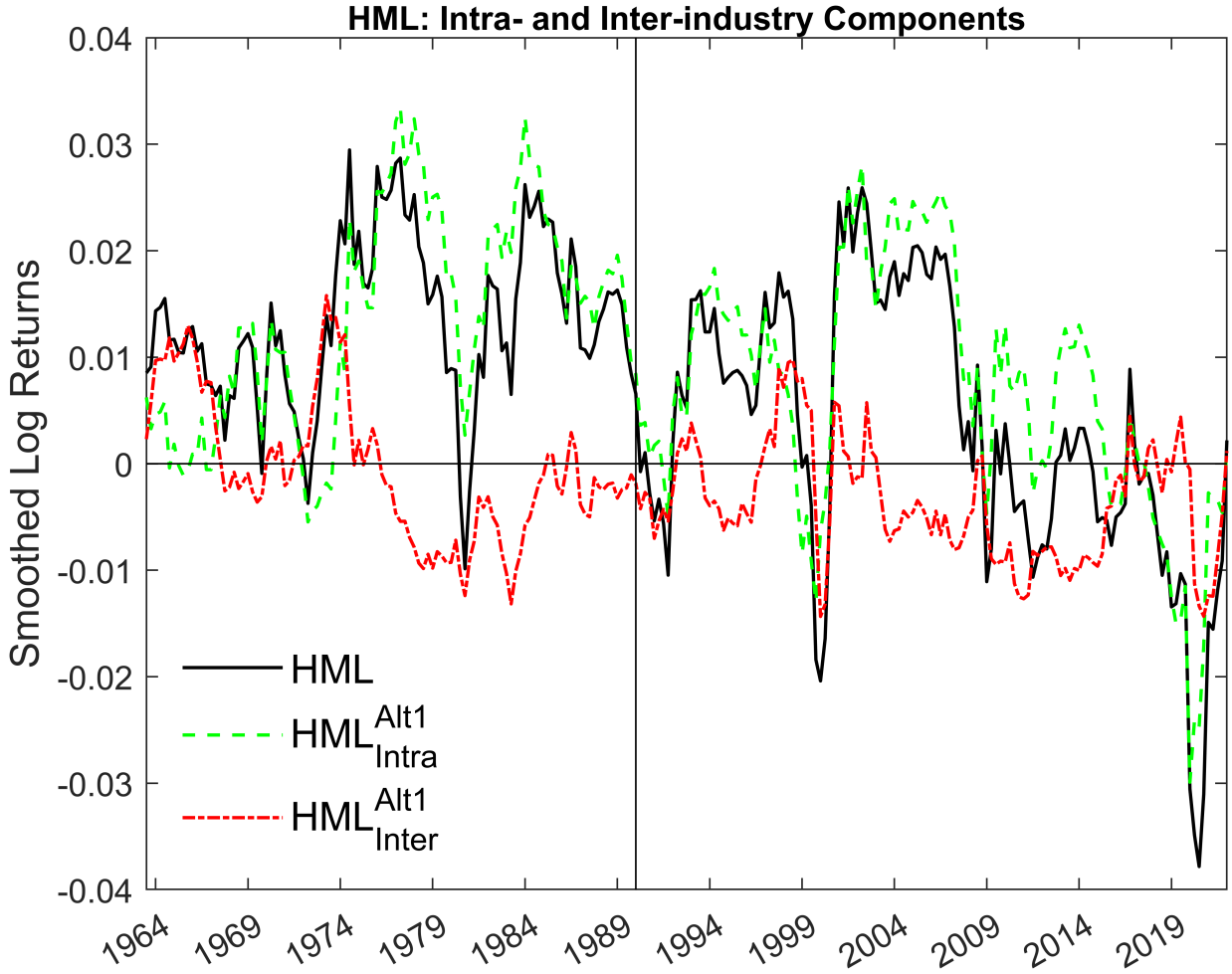


Figure 2: We plot the results from the Appendix Table 3 decomposition of HML into its intra- and inter-industry components for the Compustat period from 1963Q3-2022Q1 using the first alternative way of measuring HML’s intra-industry component. The solid black line shows the smoothed log return to Fama and French’s (1993) HML; the dashed green line shows the smoothed log return to the intra-industry component of HML; and the dashed-dotted red line shows the smoothed log return to the inter-industry component. The series are smoothed with a trailing exponentially weighted moving average in which the decay parameter is set to 0.08 per quarter, and the smoothed series is generated, for example, as $MA_t(\text{HML}) = 0.08\text{HML}_t + (1-0.08)MA_{t-1}(\text{HML})$. This decay parameter implies a half-life of two years. The vertical line indicates the start of the subperiod 1990Q2-2022Q1.

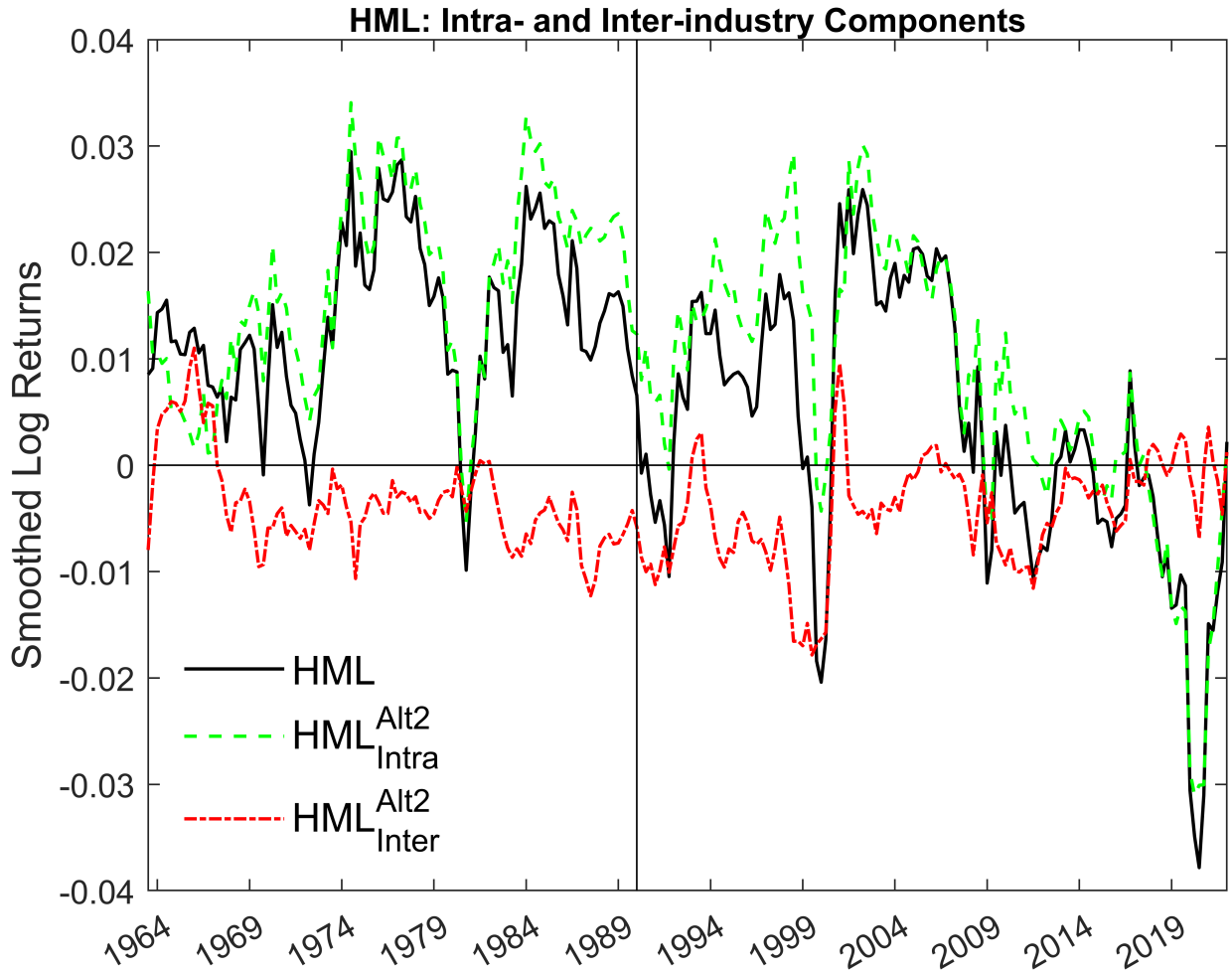


Figure 3: We plot the results from the Appendix Table 4 decomposition of HML into its intra- and inter-industry components for the Compustat period from 1963Q3-2022Q1 using the second alternative way of measuring HML’s intra-industry component. The solid black line shows the smoothed log return to Fama and French’s (1993) HML; the dashed green line shows the smoothed log return to the intra-industry component of HML; and the dashed-dotted red line shows the smoothed log return to the inter-industry component. The series are smoothed with a trailing exponentially weighted moving average in which the decay parameter is set to 0.08 per quarter, and the smoothed series is generated, for example, as $MA_t(\text{HML}) = 0.08\text{HML}_t + (1-0.08)MA_{t-1}(\text{HML})$. This decay parameter implies a half-life of two years. The vertical line indicates the start of the subperiod 1990Q2-2022Q1.

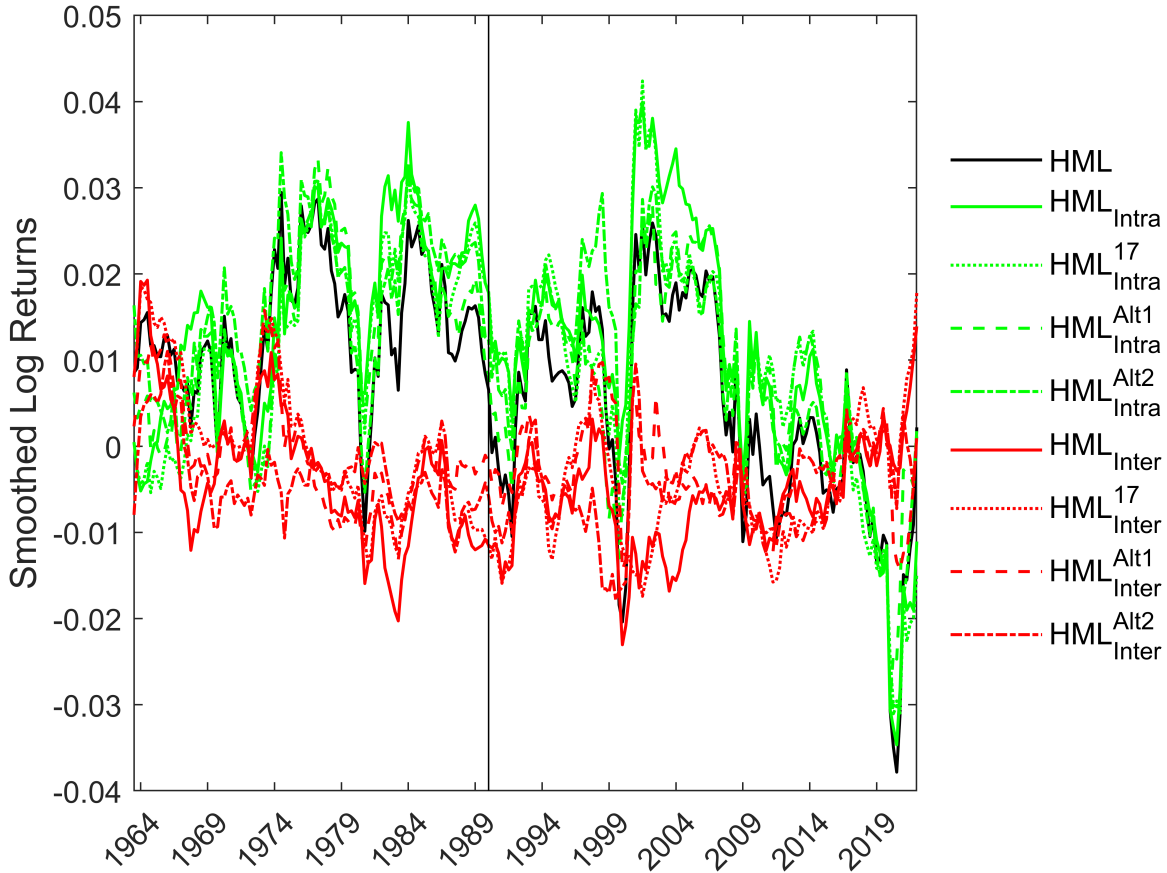


Figure 4: This figure plots the results from the four decompositions of HML into intra- and inter-industry components for the Compustat period from 1963Q3-2022Q1. The solid black line shows the smoothed log return to Fama and French’s (1993) HML; the green lines show the smoothed log return to the various intra-industry components of HML; and the red lines show the smoothed log return to the corresponding inter-industry components. The series are smoothed with a trailing exponentially weighted moving average in which the decay parameter is set to 0.08 per quarter, and the smoothed series is generated, for example, as $MA_t(\text{HML}) = 0.08\text{HML}_t + (1-0.08)MA_{t-1}(\text{HML})$. This decay parameter implies a half-life of two years. The vertical line indicates the start of the subperiod 1990Q2-2022Q1.

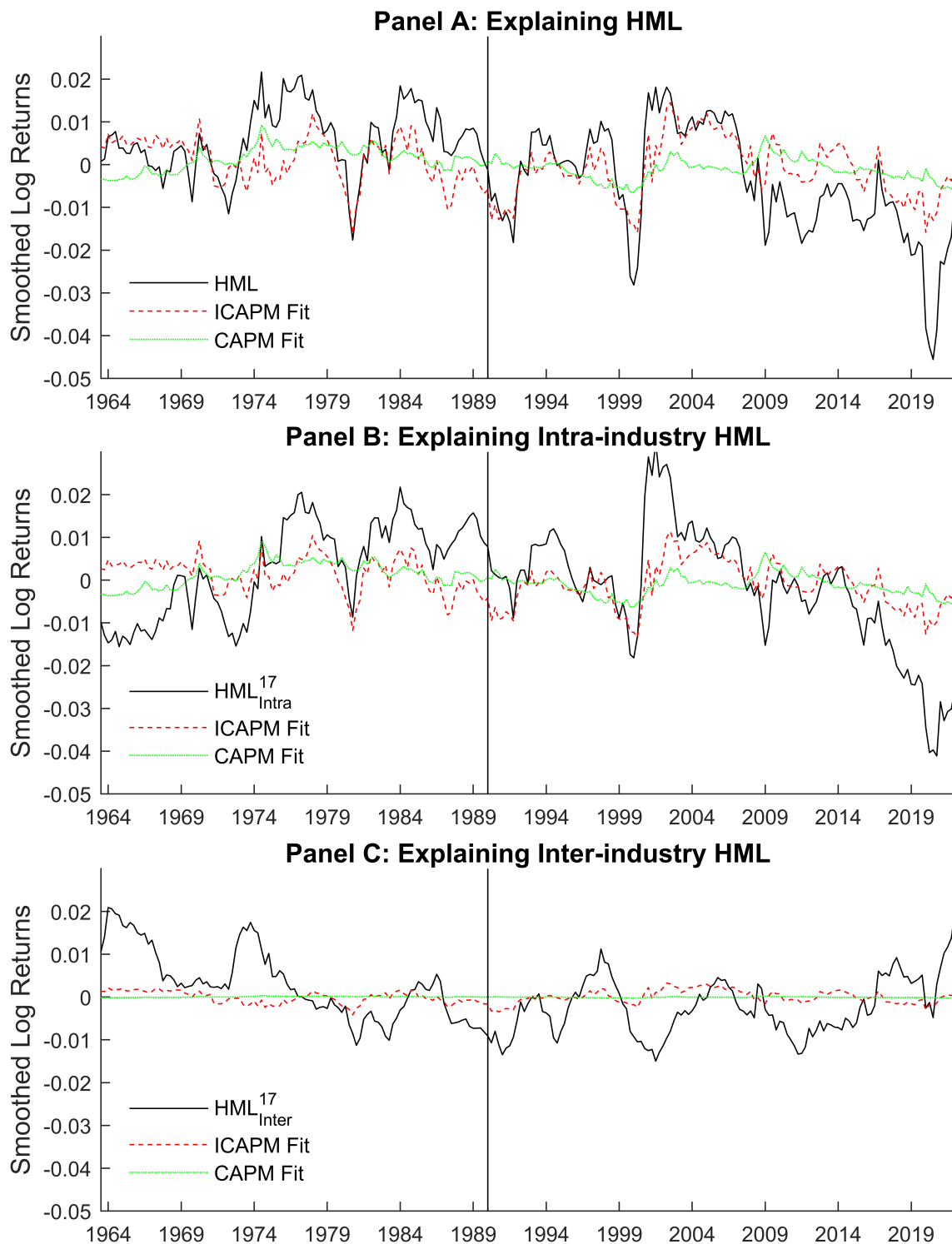


Figure 5: We explain time-series variation in demeaned HML, HML_{Intra}^{17} , and HML_{Inter}^{17} , based on Fama and French's classification of firms into 17 industries. Smoothed log returns are plotted with a solid black line while smoothed ICAPM (CAPM) fitted values are plotted with a dashed red (dotted green) line.

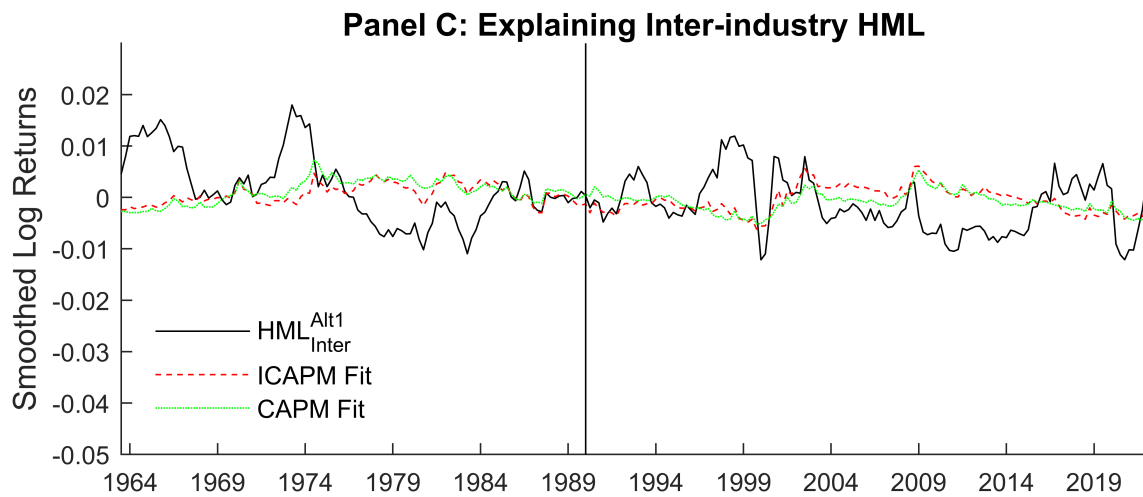
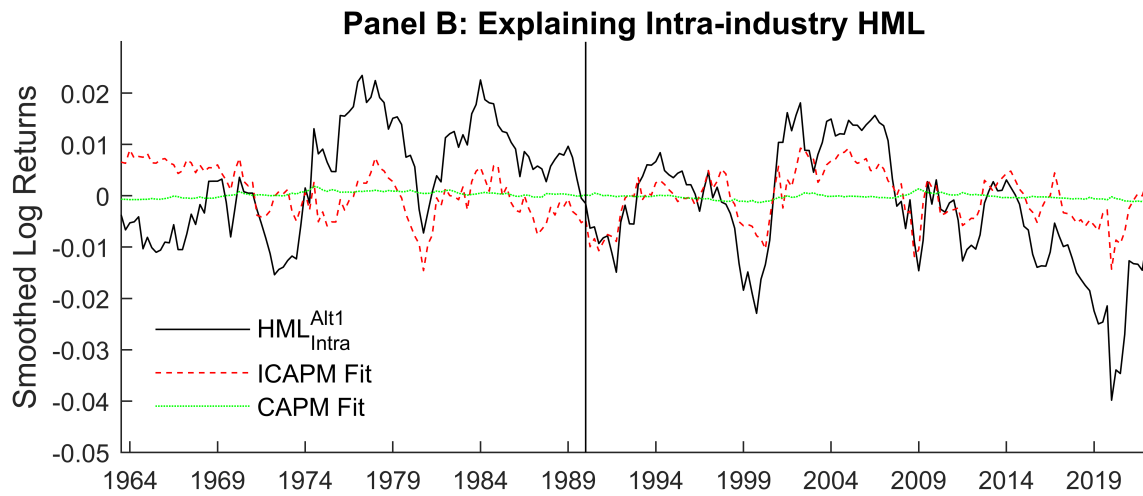
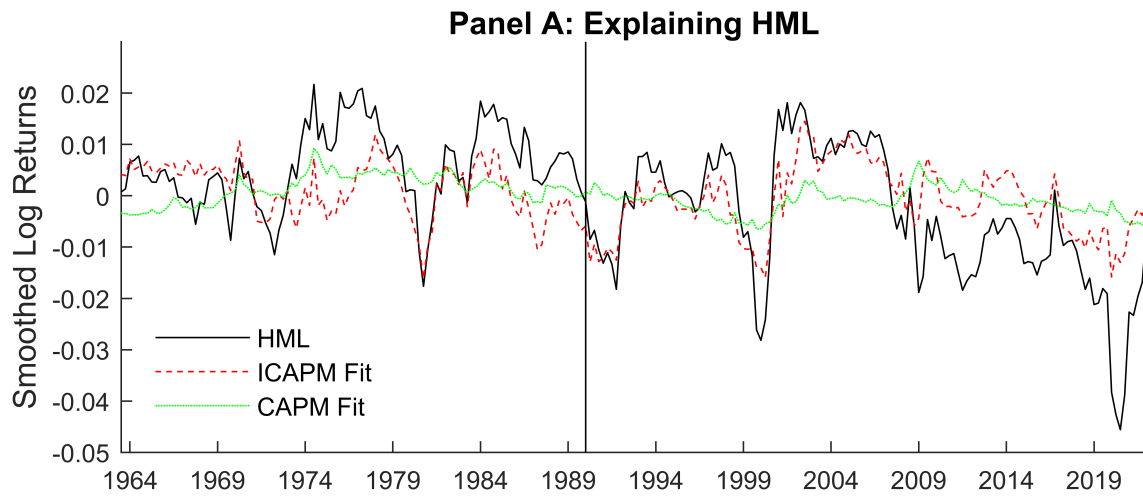


Figure 6: We explain time-series variation in demeaned HML, HML_{Intra}^{Alt1} , and HML_{Inter}^{Alt1} where we use our first alternative way of measuring HML's intra-industry component. Smoothed log returns are plotted with a solid black line while smoothed ICAPM (CAPM) fitted values are plotted with a dashed red (dotted green) line.

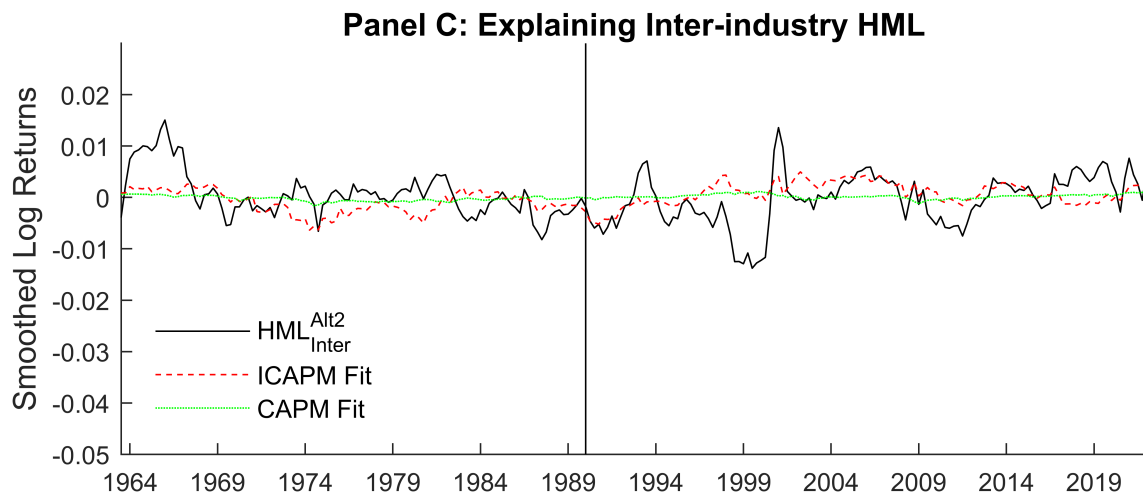
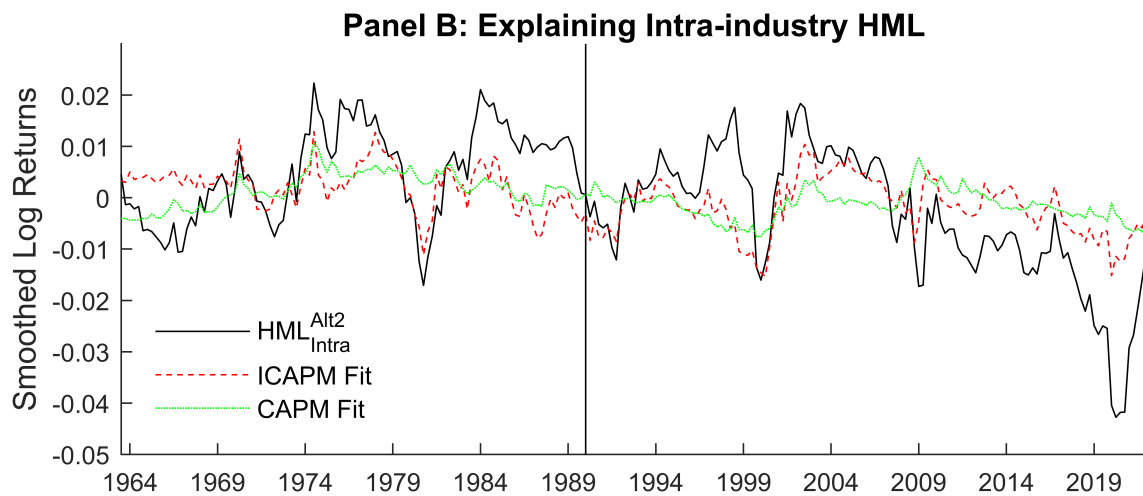
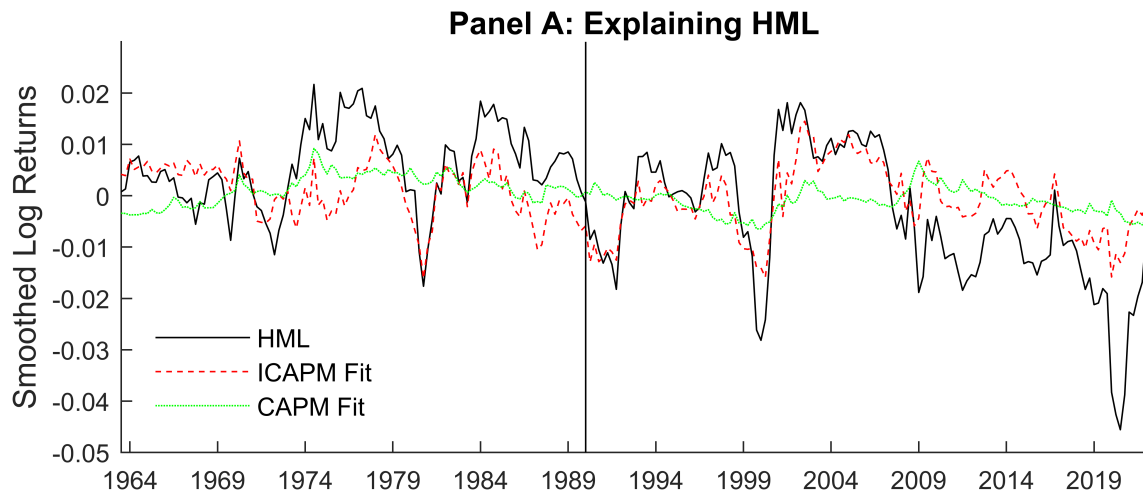


Figure 7: We explain time-series variation in demeaned HML, HML_{Intra}^{Alt2} , and HML_{Inter}^{Alt2} where we use our second alternative way of measuring HML's intra-industry component. Smoothed log returns are plotted with a solid black line while smoothed ICAPM (CAPM) fitted values are plotted with a dashed red (dotted green) line.

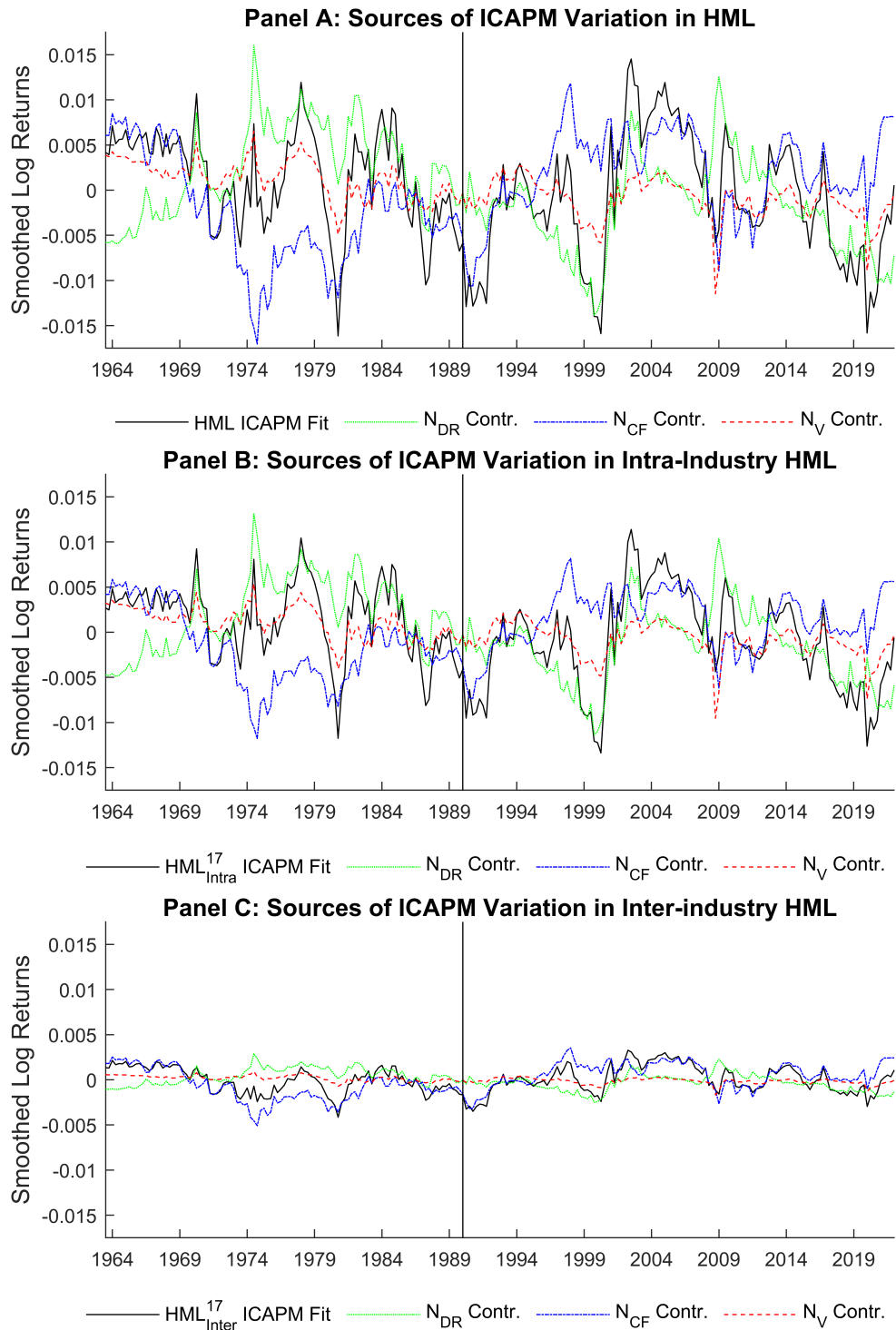


Figure 8: We plot the components of the ICAPM fit for HML, HML_{Intra}^{17} , and HML_{Inter}^{17} , based on Fama and French’s classification of firms into 17 industries. The solid black line shows the smoothed ICAPM fit; the dashed green line shows the smoothed contribution of N_{DR} ; the dashed blue line shows the smoothed contribution of N_{CF} ; and the dashed red line shows the smoothed contribution of N_V .

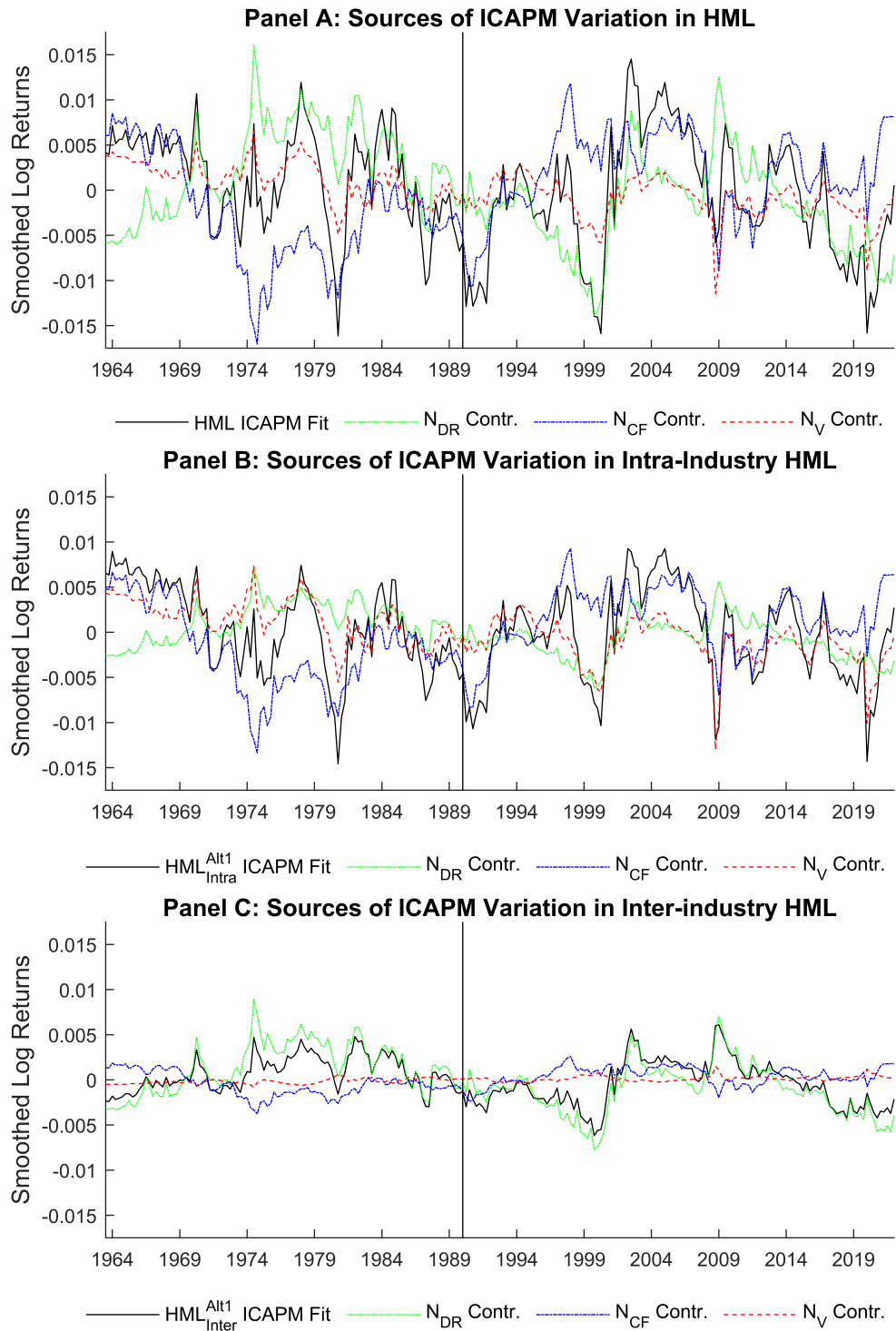


Figure 9: We plot the components of the ICAPM fit for HML, HML_{Intra}^{Alt1} , and HML_{Inter}^{Alt1} using the first alternative way of measuring HML's intra-industry component. The solid black line shows the smoothed ICAPM fit; the dashed green line shows the smoothed contribution of N_{DR} ; the dashed blue line shows the smoothed contribution of N_{CF} ; and the dashed red line shows the smoothed contribution of N_V .

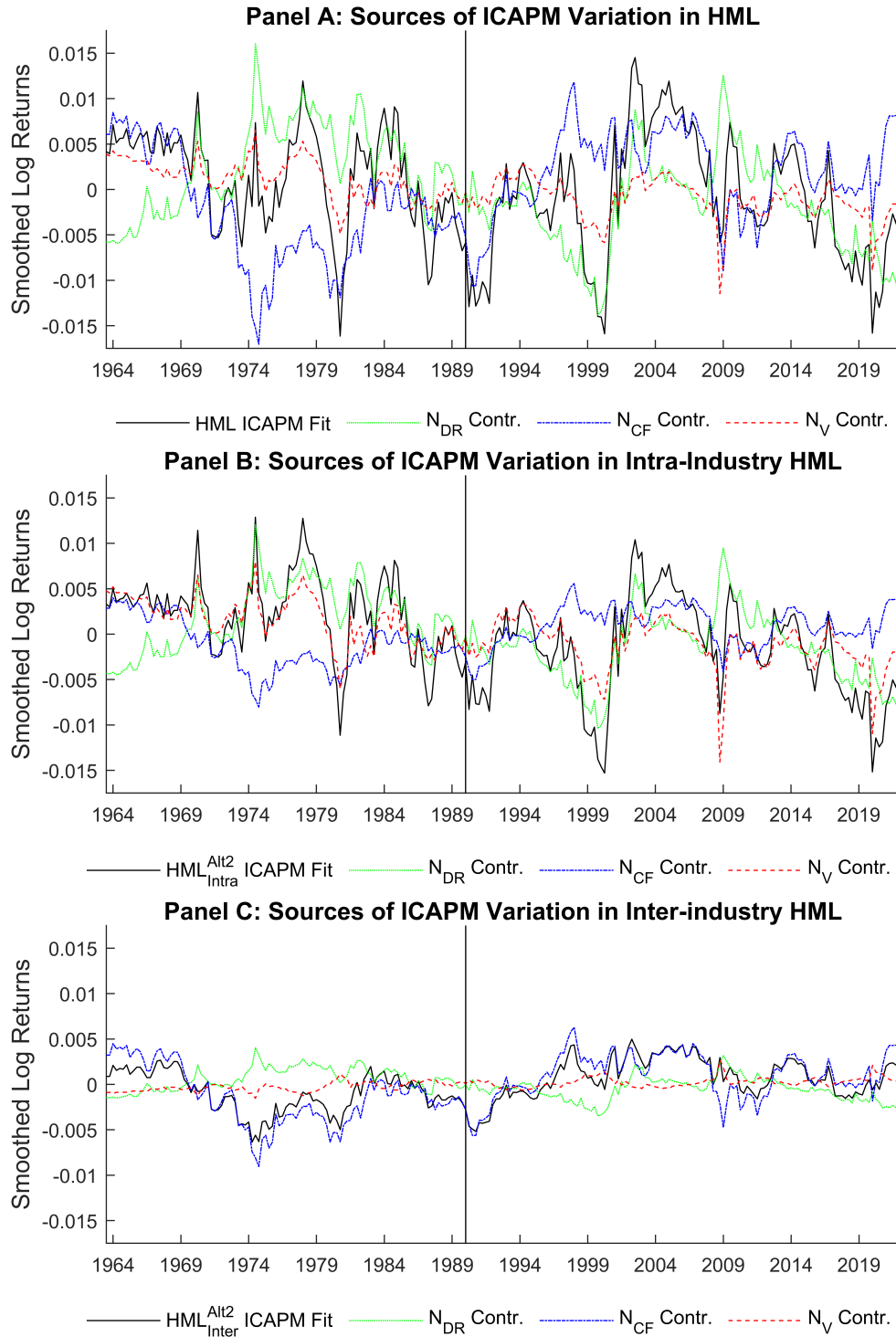


Figure 10: We plot the components of the ICAPM fit for HML, HML_{Intra}^{Alt2} , and HML_{Inter}^{Alt2} using the second alternative way of measuring HML's intra-industry component. The solid black line shows the smoothed ICAPM fit; the dashed green line shows the smoothed contribution of N_{DR} ; the dashed blue line shows the smoothed contribution of N_{CF} ; and the dashed red line shows the smoothed contribution of N_V .

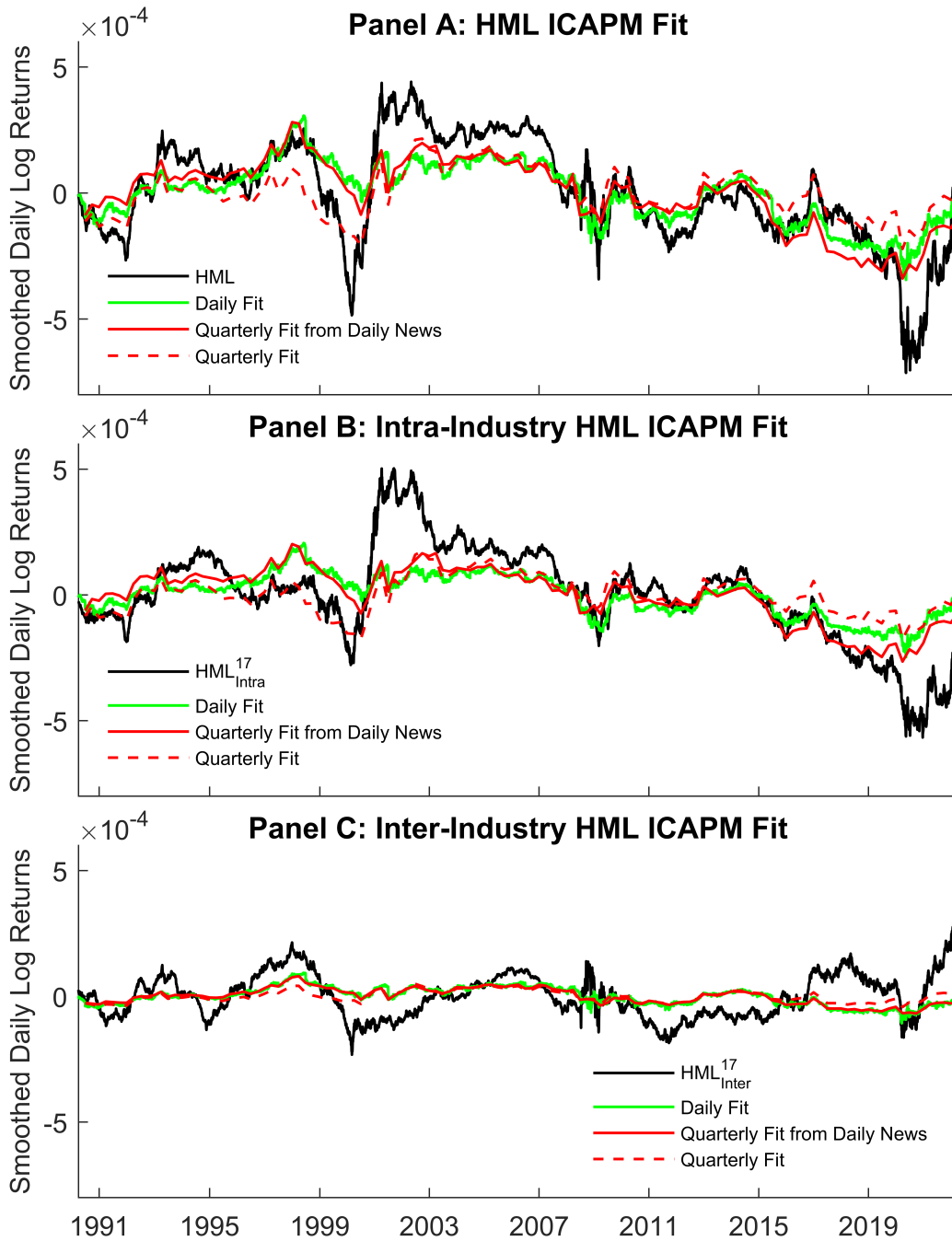


Figure 11: We plot smoothed ICAPM Fits for HML, HML_{Intra}^{17} , and HML_{Inter}^{17} (based on Fama and French's 17 industries) over the 1990-04-02 to 2022-03-31 subsample as generated by the quarterly VAR in Appendix Table 1 and daily VAR in Appendix Table 2.

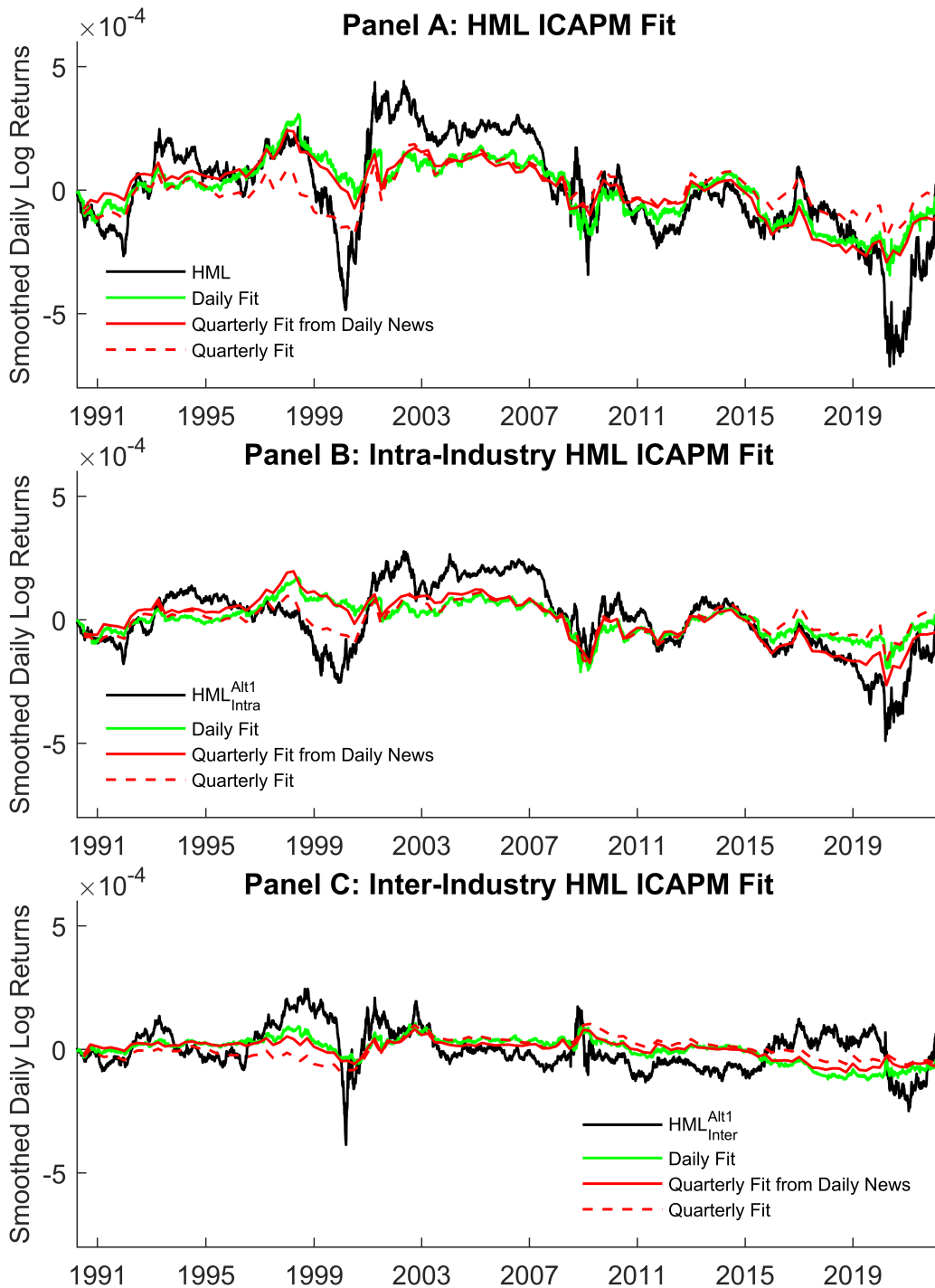


Figure 12: We plot smoothed ICAPM Fits for HML , HML_{Intra}^{Alt1} , and HML_{Inter}^{Alt1} (using the first alternative way of measuring HML's intra-industry component) over the 19900402-20220331 subsample as generated by the quarterly VAR in Appendix Table 1 and daily VAR in Appendix Table 2.

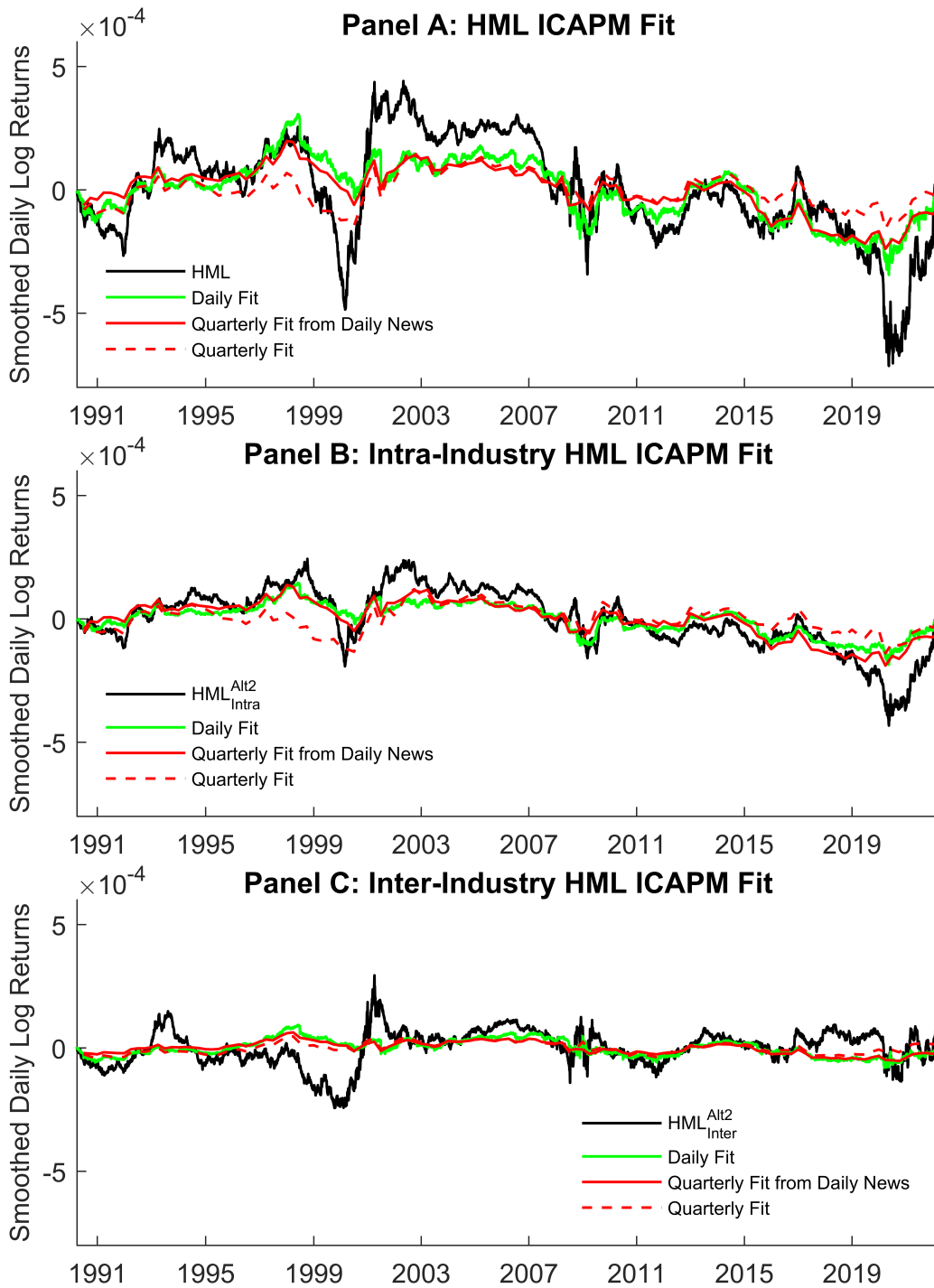


Figure 13: We plot smoothed ICAPM Fits for HML, HML_{Intra}^{Alt2} , and HML_{Inter}^{Alt2} (using the second alternative way of measuring HML's intra-industry component) over the 19900402-20220331 subsample as generated by the quarterly VAR in Appendix Table 1 and daily VAR in Appendix Table 2.

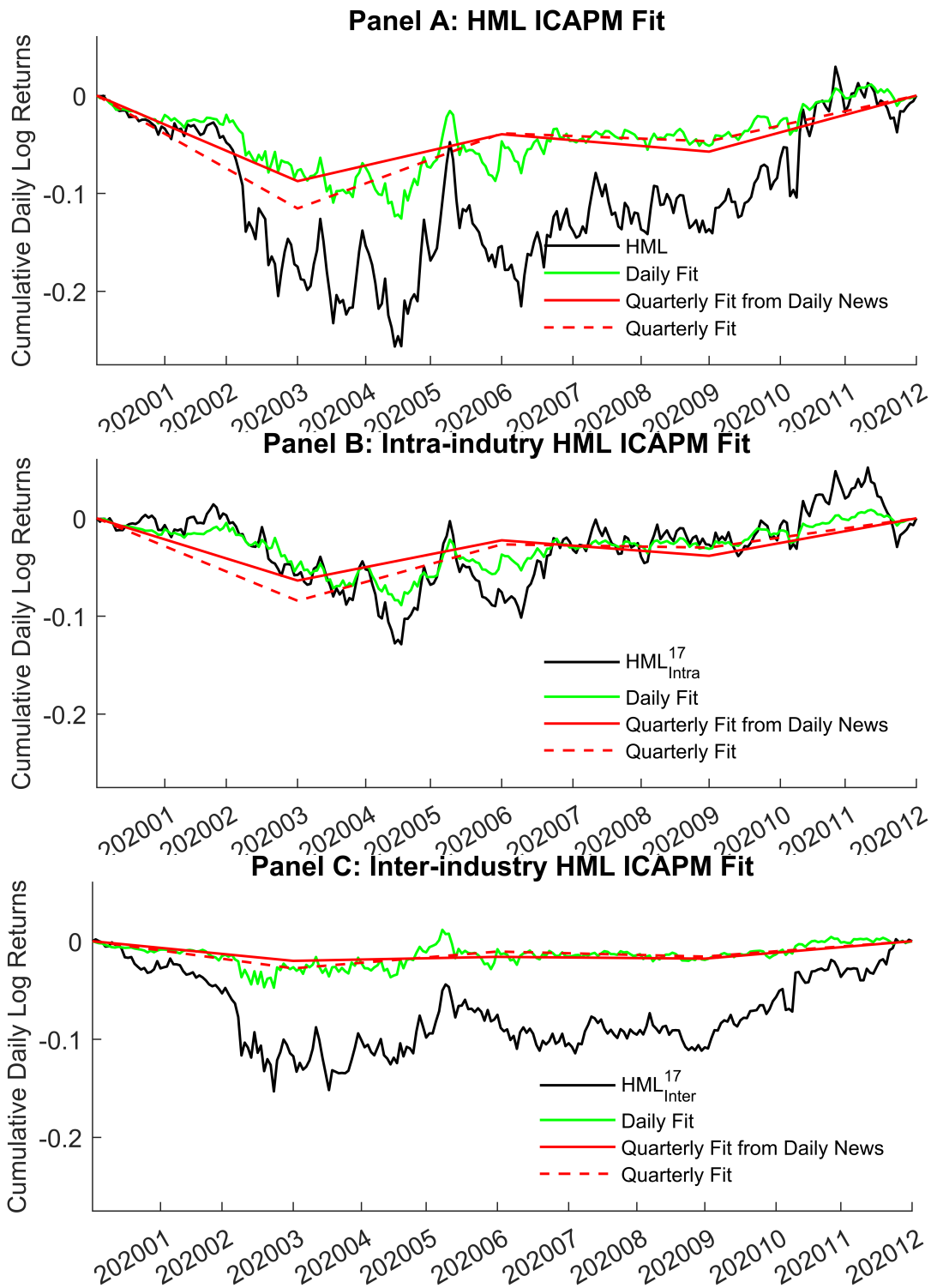


Figure 14: We plot various cumulative ICAPM fits for HML, HML_{Intra}^{17} , and HML_{Inter}^{17} (based on Fama and French's 17 industries) from the quarterly and daily VARs for the 2020-01-02 to 2020-12-31 subsample.

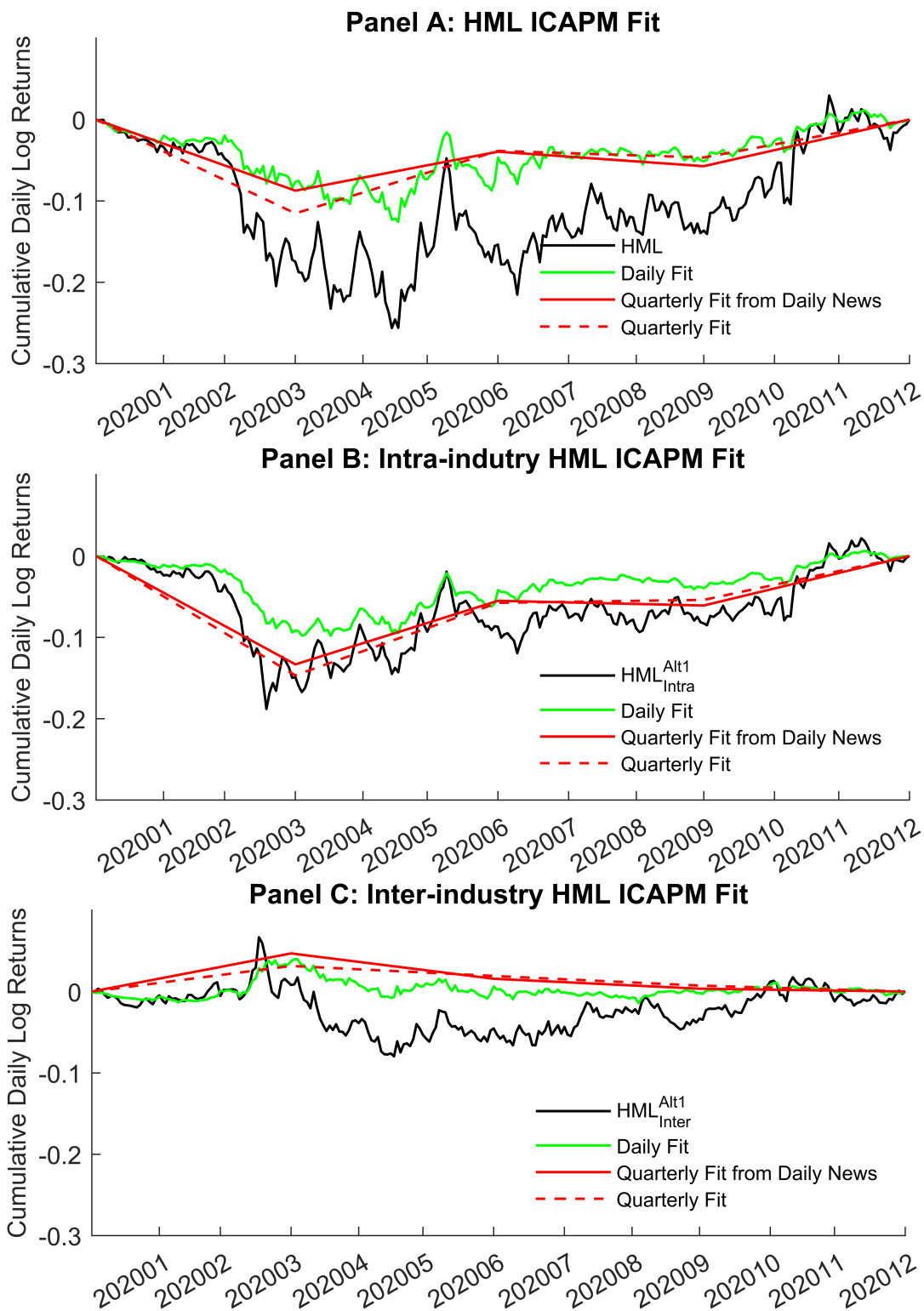


Figure 15: We plot various cumulative ICAPM fits for HML, HML_{Intra}^{Alt1} , and HML_{Inter}^{Alt1} (using the first alternative way of measuring HML's intra-industry component) from the quarterly and daily VARs for the 20200101-20201231 subsample.

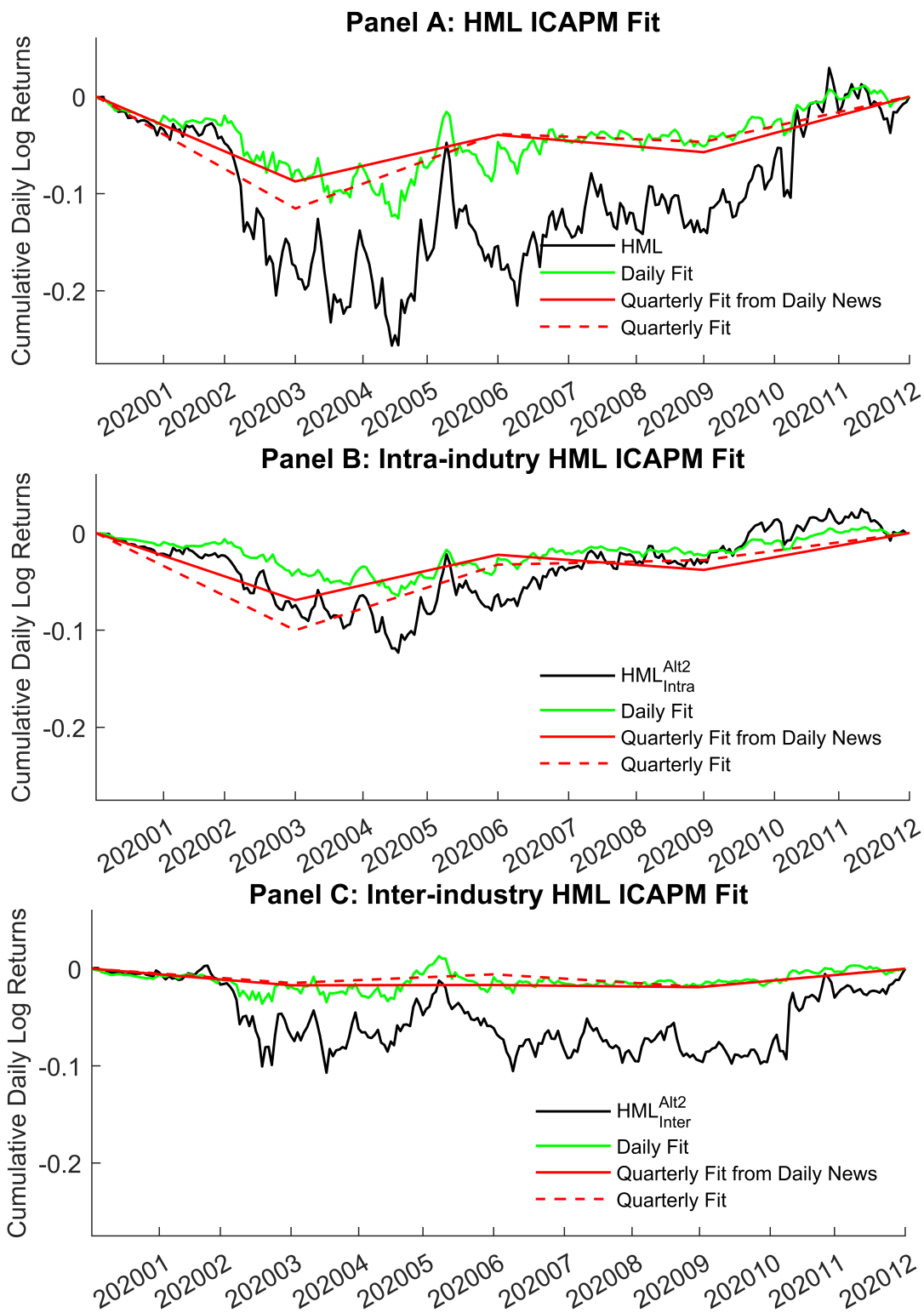


Figure 16: We plot various cumulative ICAPM fits for HML, HML_{Intra}^{Alt2} , and HML_{Inter}^{Alt2} (using the second alternative way of measuring HML's intra-industry component) from the quarterly and daily VARs for the 20200101-20201231 subsample.

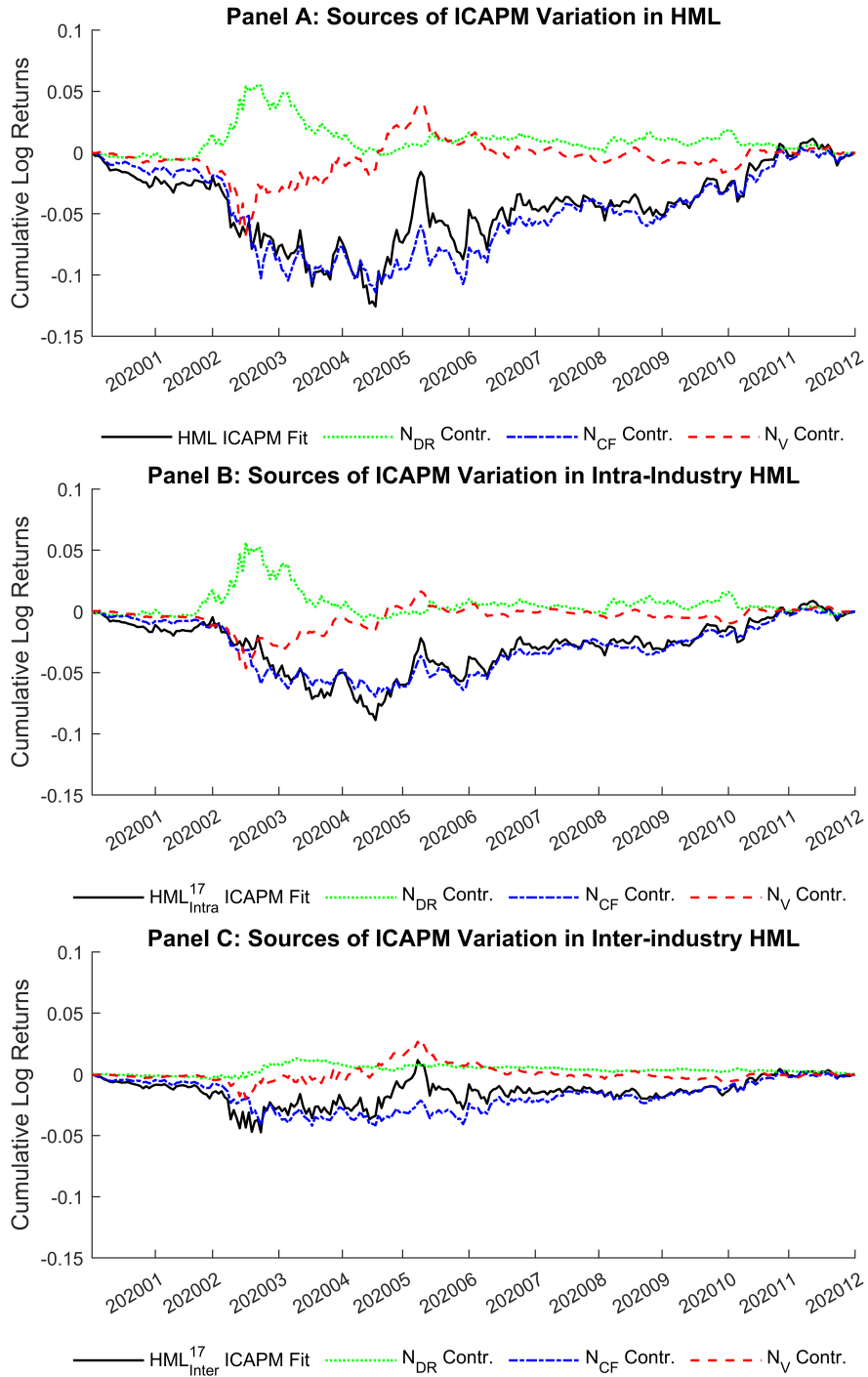


Figure 17: We plot the components of the ICAPM fit for HML, HML_{Intra}^{17} , and HML_{Inter}^{17} (based on Fama and French’s 17 industries) for the 2020-01-02 to 2020-12-31 subsample. The solid black line shows the smoothed ICAPM fit; the dashed green line shows the smoothed contribution of N_{DR} ; the dashed blue line shows the smoothed contribution of N_{CF} ; and the dashed red line shows the smoothed contribution of N_V .

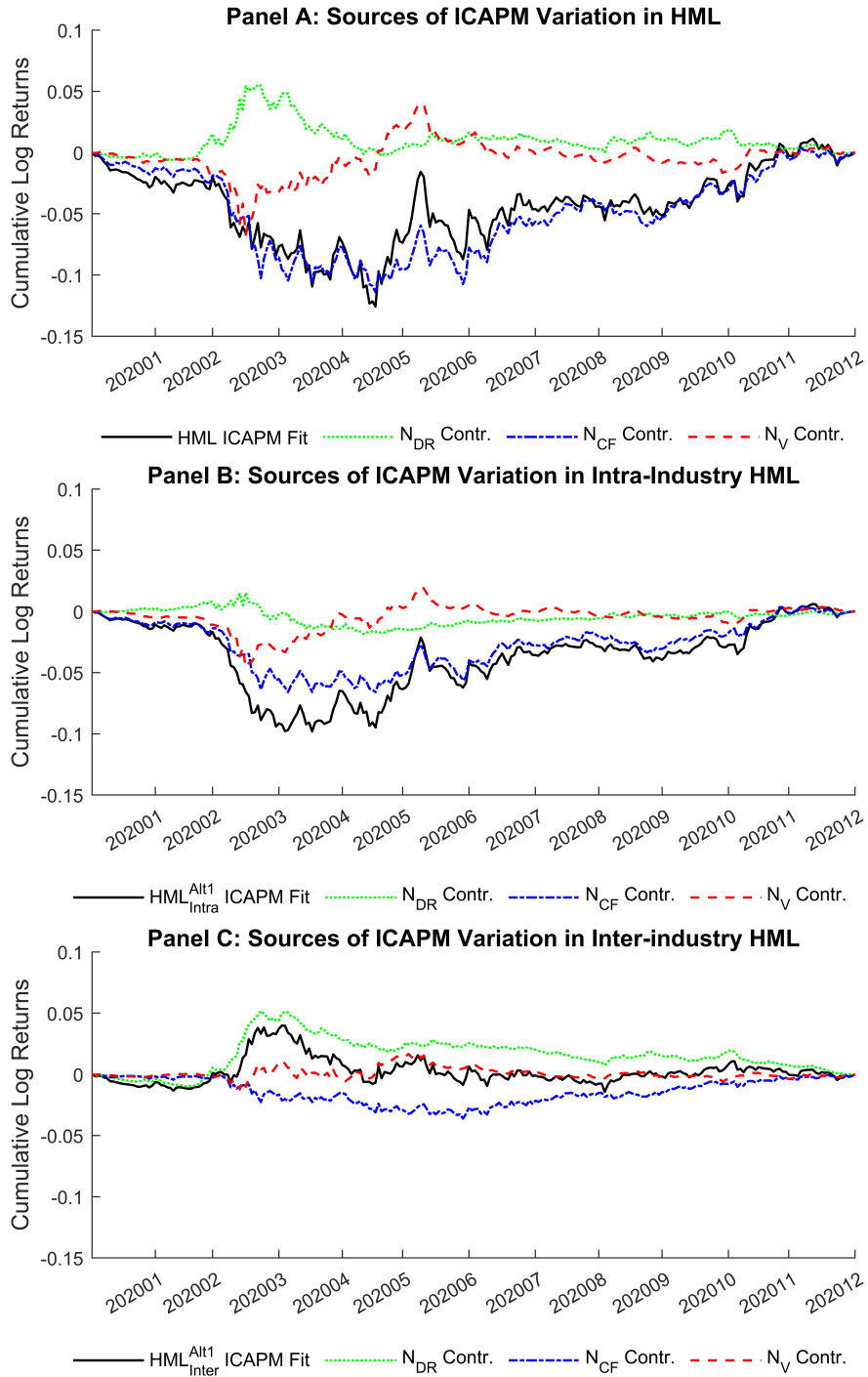


Figure 18: We plot the components of the ICAPM fit for HML, HML_{Intra}^{Alt1} , and HML_{Inter}^{Alt1} (using the first alternative way of measuring HML's intra-industry component) for the 20200101-20201231 subsample. The solid black line shows the smoothed ICAPM fit; the dashed green line shows the smoothed contribution of N_{DR} ; the dashed blue line shows the smoothed contribution of N_{CF} ; and the dashed red line shows the smoothed contribution of N_V .

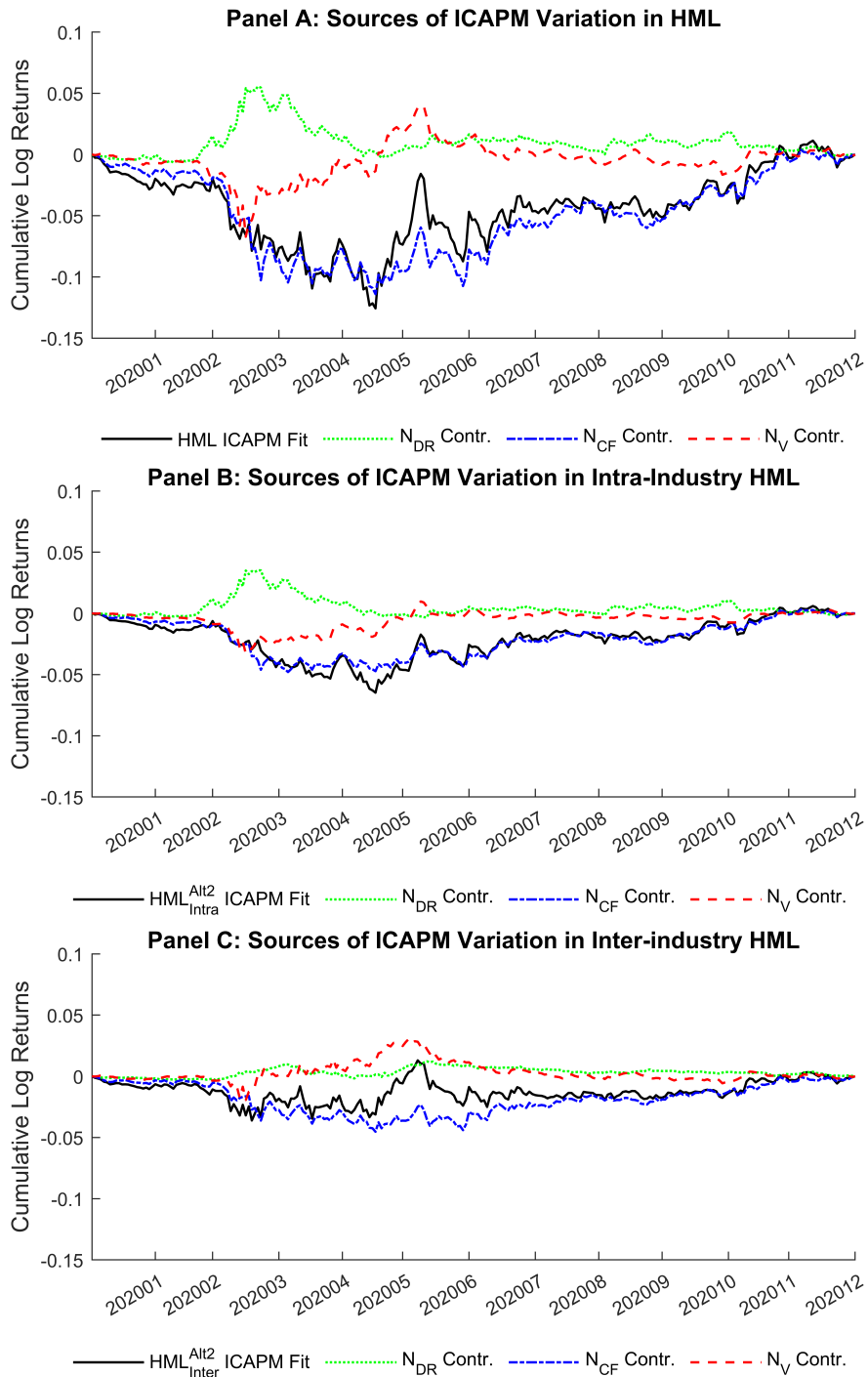


Figure 19: We plot the components of the ICAPM fit for HML, HML_{Intra}^{Alt2} , and HML_{Inter}^{Alt2} (using the second alternative way of measuring HML's intra-industry component) for the 20200101-20201231 subsample. The solid black line shows the smoothed ICAPM fit; the dashed green line shows the smoothed contribution of N_{DR} ; the dashed blue line shows the smoothed contribution of N_{CF} ; and the dashed red line shows the smoothed contribution of N_V .

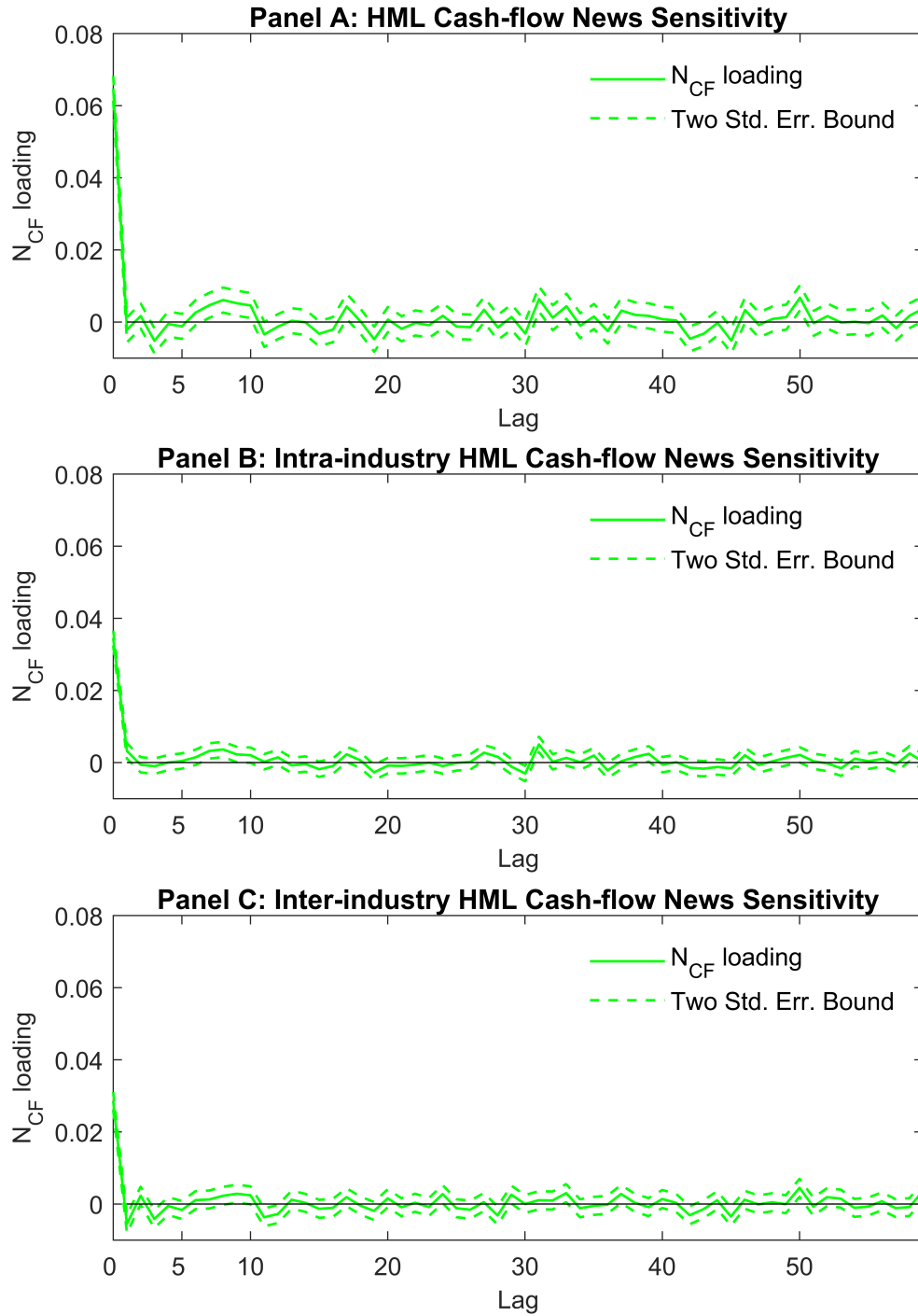


Figure 20: This figure shows the coefficients on contemporaneous and lagged cash-flow news that correspond to the daily ICAPM regressions in Table 5 of the main paper.

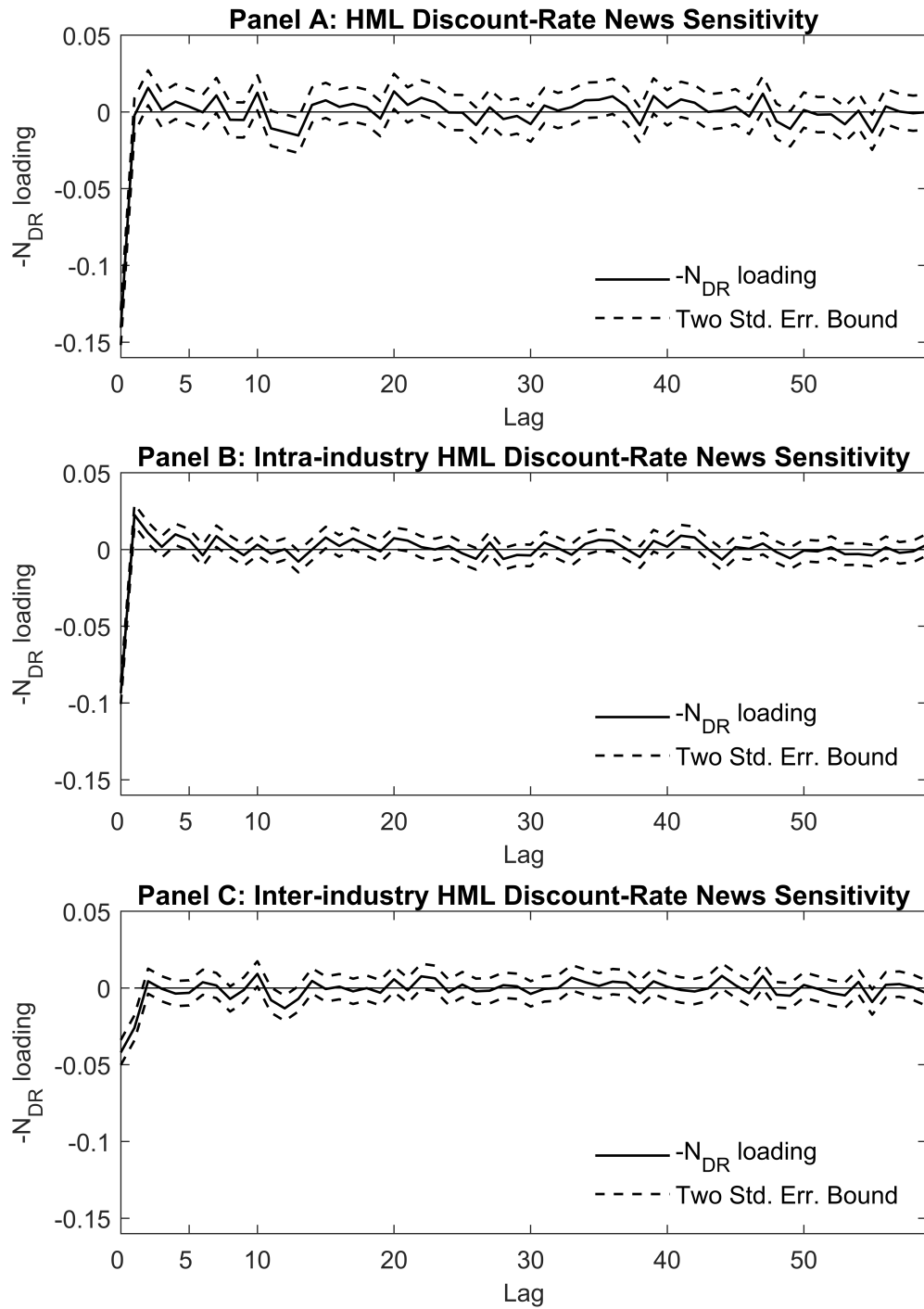


Figure 21: This figure shows the coefficients on contemporaneous and lagged negative discount-rate news that correspond to the daily ICAPM regressions in Table 5 of the main paper.

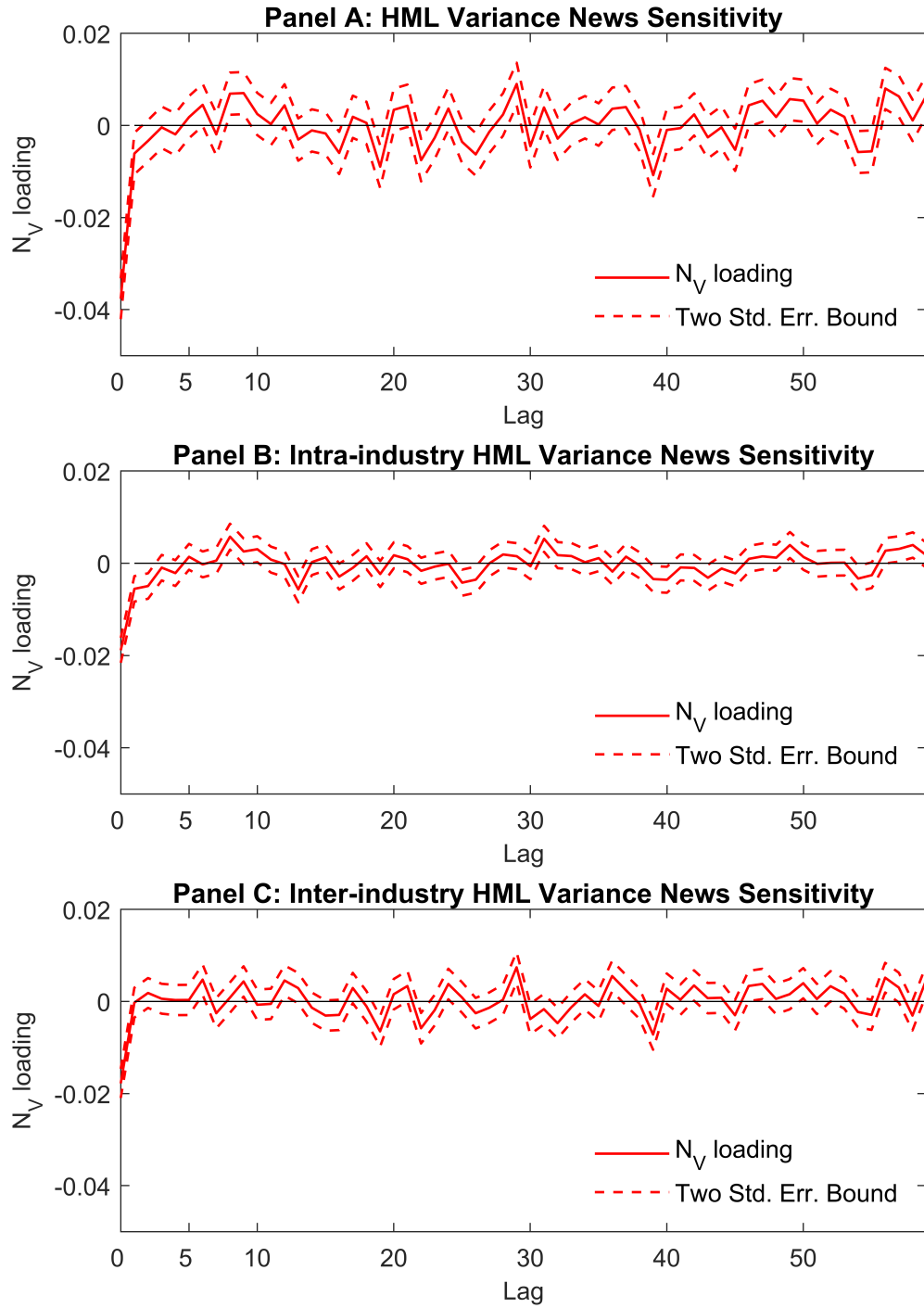


Figure 22: This figure shows the coefficients on contemporaneous and lagged variance news that correspond to the daily ICAPM regressions in Table 5 of the main paper.