### **PROJECT IDEAS**

#### **Petnica Science Center**

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## **Topics**

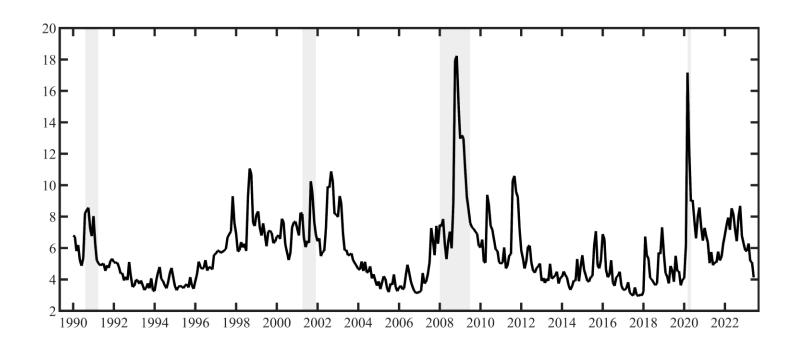
- VIX-based dynamic hedging
- Value vs. growth: a macro-based strategy
- Stock return kurtosis and crash exposure
- Timing beta exposure using idiosyncratic volatility
- Cross-sectional volatility spread as a timing signal
- Testing and extending the Betting Against Beta strategy
- Exploiting predictable intraday patterns with 0DTE options



## VIX-BASED DYNAMIC HEDGING

## Idea

Dynamically hedge downside risk in an equity portfolio





# Objective

- Volatility timing relies on scaling exposure
  - Barroso & Santa-Clara (2015), Daniel & Moskowitz (2016), Moreira & Muir (2017), Liu et al. (2019), Cederburg et al. (2020), Eisdorfer & Misirli (2020), Barroso & Detzel (2021), Wang & Yan (2021)
- Use the VIX as a forward-looking signal
  - Božović (2024)





- Trigger signal (VIX threshold or regime classification)
  - Define volatility regimes using VIX:

Low: VIX < 15</li>

Medium: 15 ≤ VIX < 25</li>

• High: VIX ≥ 25

- Alternatively, use a percentile-based threshold (e.g., hedge when VIX is in the top 20% of its trailing 1-year distribution)
- Use liquid, high-convexity hedging instruments



#### Hedge construction rules

- When VIX enters a high regime:
  - Buy out-of-the-money (OTM) SPX put options (e.g., 5% OTM, 1-month expiry)
  - Size the notional of puts to hedge a portion of portfolio delta (e.g., hedge 20–50% of the equity notional depending on VIX level)
  - Alternatively, buy VIX calls or long VIX futures to benefit directly from a spike in volatility
- When VIX is low:
  - Let the hedges expire, or monetize residual value if they're in the money
  - Refrain from initiating new hedges



#### Dynamic adjustment

- Rebalance hedge positions weekly or daily, based on updated VIX levels
- Use rolling windows (e.g., trailing 21-day average VIX) for smoothing



## Data

- VIX
- Hedging instruments
  - SPX puts
  - VIX futures or VIX calls
- More feasible ETF equivalents:
  - VIXY
  - SPLV
  - TLT
  - SH
  - SWAN, TAIL



## ETF-based hedge execution

 When VIX triggers a high-risk regime, execute one of the following:

ETF Ticker	Description	Role in Hedge
VIXY	Short-term VIX futures ETF	Direct exposure to volatility spikes
SPLV	S&P 500 Low Volatility ETF	Reduce portfolio beta, smooth drawdowns
TLT	Long-term Treasuries (20+ yr)	Flight-to-quality hedge
SH	Short S&P 500 (inverse ETF, 1x)	Linear hedge vs. equity downside



## Implementation example

- Assume a fund that manages a \$100M active equity portfolio (pick any core strategy)
- In high-VIX regime:
  - Reduce 5–10% equity exposure
  - Reallocate into a hedge basket (e.g., 50% VIXY, 25% TLT, 25% SH)
  - Keep the core strategy unchanged: this is a temporary allocation shift, not a full regime switch
- Rebalance weekly or as VIX exits the high-risk zone



## Performance expectations: derivatives

- VIX contains forwardlooking information about volatility and kurtosis
  - Not just second moments but higher-order (tail) risk
- When VIX is high:
  - Options are expensive, but so is the risk
  - You're paying for insurance when you need it
- This strategy targets the source of tail risk instead of exposure reduction





## Performance expectations: ETFs

- When markets are stable:
  - Hedge underperforms
  - But cost is limited
- During tail events:
  - Hedge ETFs (esp. VIXY and TLT) gain sharply
- Net result:
  - Smoothed equity P&L
  - Improved Sharpe ratio
  - Lower drawdown



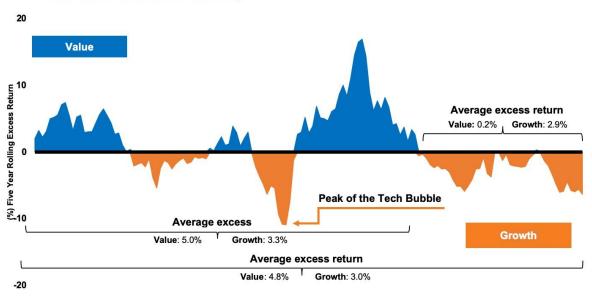
# VALUE VS. GROWTH: A MACRO-BASED STRATEGY

## Idea

 Test whether simple macro indicators can improve the performance of value-vs-growth portfolios

#### Value vs. growth

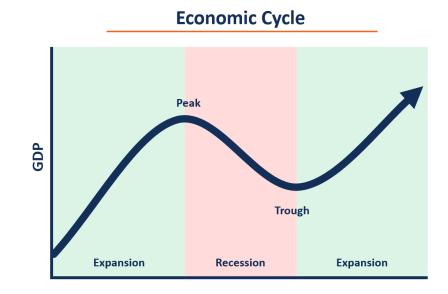
Will value re-take leadership?





# Objective

- Dynamically allocate between value and growth equities based on prevailing macroeconomic conditions
- Exploit predictable shifts in factor performance across business cycles
- Generate improved riskadjusted returns by aligning exposures with the macro environment



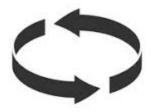


#### Signal construction:

- Define macro regimes (e.g., expansion, slowdown, stagflation, recovery) using thresholds
- Estimate expected relative performance of value vs. growth in each regime

#### Portfolio implementation:

- Allocate between value and growth based on current macro regime
- Adjust weights monthly or quarterly with a lag to avoid data snooping





### Data

- Monthly macroeconomic indicators (e.g., inflation, unemployment, real GDP growth)
- Fama-French style portfolios



## Performance expectations

#### • Why it works?

- Value and growth styles respond differently to inflation, interest rates and earnings outlooks
- Timing exposure reduces prolonged underperformance from static allocation
- Macro regimes exhibit persistence, providing exploitable predictability

#### Expected outcomes:

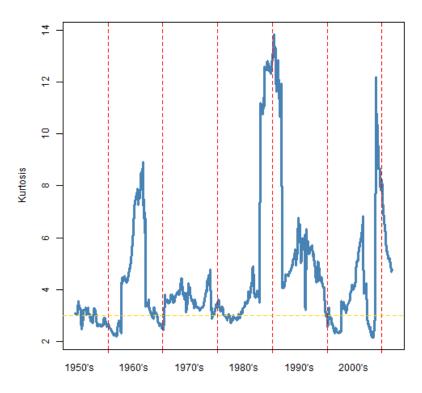
- Higher Sharpe ratio than unconditional value or growth exposure
- Reduced drawdowns during adverse cycles
- Improved performance over full market cycles



# STOCK RETURN KURTOSIS AND CRASH EXPOSURE

## Idea

 Investigate whether elevated realized kurtosis in index or stock returns signals impending market crashes or volatility spikes





# Objective

- Identify and measure individual stock exposure to tail risk via return kurtosis
- Construct portfolios that balance return and crash exposure by filtering high-kurtosis stocks
- Explore the pricing of higher-moment risk in crosssectional stock returns



#### Signal construction:

- Estimate ex-ante return kurtosis for each stock
- Identify stocks with persistently high crash risk profiles
- Form long-short portfolios that overweight low-kurtosis stocks and underweight high-kurtosis stocks

#### Portfolio implementation:

- Simple equal weighting
- Monthly rebalancing to update crash exposure estimates



#### Data

- Daily or weekly stock returns (e.g., CRSP data, S&P 500 universe)
- Computed rolling kurtosis over fixed windows (e.g., 6 or 12 months)



## Performance expectations

#### Why it works?

- Kurtosis captures crash risk that volatility overlooks
- Market participants underprice extreme left-tail risk, creating a premium for low-kurtosis stocks
- Avoiding crash-prone assets enhances compound returns during market stress

#### Expected outcomes:

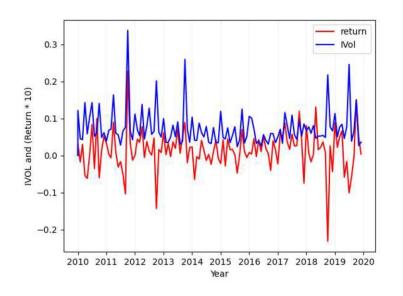
- Improved downside protection and lower tail risk exposure
- Consistent performance during crisis periods and high-volatility regimes
- Potential alpha from neglected pricing of higher-moment risks

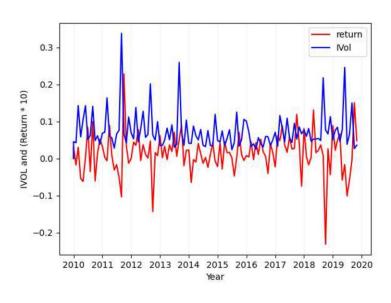


# TIMING BETA EXPOSURE USING IDIOSYNCRATIC VOLATILITY

## Idea

- Test whether periods of high average idiosyncratic volatility predict lower equity market returns
- Use that to modulate beta exposure







# Objective

- Investigate whether idiosyncratic volatility (IVOL) contains information about future stock returns
- Dynamically time market beta exposure based on aggregate IVOL levels
- Design a risk-managed portfolio that adapts to changes in firm-specific uncertainty
- Explore IVOL as a proxy for sentiment or market fragility



#### Signal construction:

- Cross-sectionally rank stocks by IVOL
- Define IVOL quintiles or deciles
- Compute average market beta for each group

#### Timing rule:

- When aggregate IVOL is high (market-wide average IVOL above threshold):
  - Reduce beta exposure (shift to low-beta, low-IVOL stocks)
- When aggregate IVOL is low:
  - Increase beta exposure (allow for higher-beta positions)
- This can be applied to dynamically allocate weights in a betaneutral or risk-managed portfolio.



## Data

- Daily returns of individual stocks or portfolios
- Market model regressions to extract IVOL (residual variance from CAPM)
- Aggregate IVOL signal constructed by averaging across stocks or sectors



## Performance expectations

#### • Why it works?

- IVOL reflects stock-specific noise or disagreement among investors
- High IVOL periods are associated with lower aggregate returns, especially for high-beta stocks
- Using IVOL helps time factor exposures more precisely, especially when traditional beta strategies underperform

#### Expected outcomes:

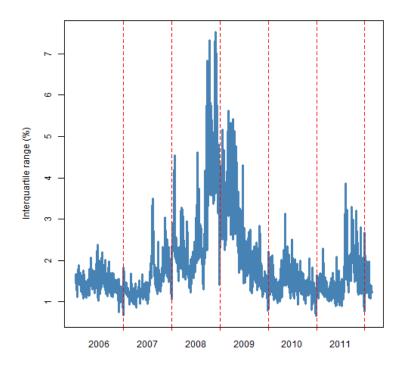
- Improved risk-adjusted returns vs. static beta exposure
- Reduced drawdowns in volatile environments
- Outperformance during market reversals or sentiment shifts



# CROSS-SECTIONAL VOLATILITY SPREAD AS A TIMING SIGNAL

### Idea

 Use the spread between the most and least volatile stocks as a signal for aggregate market volatility or trend reversals





# Objective

- Explore the predictive power of the cross-sectional volatility spread (CSVS) for market returns
- Use dispersion in stock-level volatilities as a forward-looking risk signal
- Develop a dynamic asset allocation strategy that reduces exposure during volatility clustering
- Investigate CSVS as an early warning indicator for systemic stress or market turning points



#### Signal construction:

- Compute rolling realized volatility (e.g., 20-day or 60-day standard deviation of daily returns) for each stock
- Sort stocks by realized volatility and calculate:

$$CSVS_t = Average Vol_{top decile} - Average Vol_{bottom decile}$$

Normalize the CSVS signal:

$$z_t = \frac{\text{CSVS}_t - \mu}{\sigma}$$



#### Portfolio implementation:

- High dispersion (z > 1)
  - Signals elevated risk aversion or fragmentation
  - Defensive regime (reduce the risk exposure / beta)
- Neutral zone (-1 < z < 1)</li>
  - Signals intermediate risk aversion or steady market
  - Neutral regime (maintain the baseline risk exposure / beta)
- Low dispersion (z < −1)</li>
  - Signals low risk aversion or market calm
  - Aggressive regime (increase the risk exposure / beta)





### Data

- Daily returns of a large stock universe
- Rolling estimates of individual stock volatilities



## Performance expectations

#### • Why it works?

- Rising dispersion often precedes market corrections
- Traditional risk metrics may lag
- CSVS provides early signal of regime shifts
- Helps mitigate downside risk while exploiting calmer trends

#### Expected outcomes:

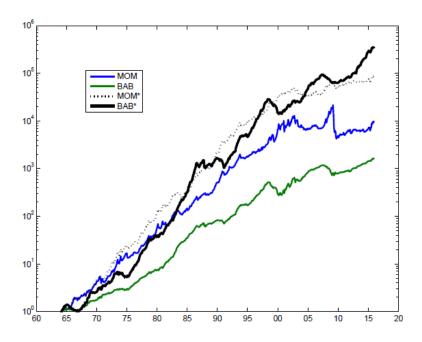
- Downside protection, superior to static allocation
- Enhanced Sharpe ratio through volatility timing
- Robust outperformance in high-volatility regimes



## TESTING AND EXTENDING THE BETTING AGAINST BETA STRATEGY

## Idea

- Replicate the classic BAB strategy using stock-level data
- Investigate whether its performance varies across volatility, interest rate or macroeconomic regimes





## Objective

- Reproduce the classic BAB strategy of Frazzini and Pedersen (2014) using updated data
- Test the robustness of BAB returns across market regimes and subperiods
- Explore enhancements by conditioning beta exposure on market volatility or macro indicators
- Investigate whether dynamic beta timing improves the risk-adjusted performance of the BAB portfolio



## Strategy outline

#### Signal construction:

- Rank stocks based on beta estimates
- Go long low-beta stocks and short high-beta stocks, dollar-neutral
- Optional enhancements:
  - Add volatility timing à la Barroso et al. (2025)
  - Add macro filters

#### Portfolio implementation:

- Monthly rebalancing
- Test performance across subperiods, sectors and market conditions

Company	♦ Symbol ♦	Zacks ▲ Rank	Zacks Rank 1-wk Ago	EPS Estimate (Current Yr)	EPS Suprise (Last Qtr)	Repor
SMITH & WE	SWHC	1	2	\$1.35	33.33%	V
ARCTIC CAT	ACAT	2	3	\$3.28	-12.82%	V
CALLAWAY G	ELY	2	3	\$0.22	35.71%	V
POLARIS IN	PII	2	3	\$6.67	1.86%	V
POOL CORP	POOL	2	4	\$2.39	-6.85%	V
BLACK DIAM	BDE	3	4	\$0.10	-15.38%	V
BRUNSWICK	BC	3	4	\$2.61	11.32%	V
MARINE PRO	MPX	3	4	\$0.23	-16.67%	V
STURM RUGE	RGR	3	1	\$4.19	19.01%	V
WEST MARIN	WMAR	3	3	\$0.63	-8.57%	NA



## Extensions considered

#### 1. Factor conditioning

- Combine BAB with cross-sectional signals like:
  - Value (e.g., low-beta + cheap stocks)
  - Quality (e.g., low-beta + high profitability)
  - Volatility (e.g., low-beta + low-vol)

#### 2. Dynamic exposure scaling

- Scale the BAB exposure based on macro indicators:
  - Credit spreads
  - VIX
  - Market trend/momentum
  - Interest rate levels

#### 3. Sector-neutral BAB

- Prevent unintended sector tilts (e.g., low-beta tends to overweight utilities)
- Apply BAB ranking within sectors, then aggregate across sectors for neutrality



### Data

- Daily or monthly returns of individual stocks
- Estimated stock betas via rolling regressions against the market portfolio (Kenneth French's Data Library)



## Performance expectations

#### • Why it works?

- Investors often favor high-beta stocks expecting higher returns, leading to their overpricing
- Low-beta stocks can be underpriced, offering superior risk-adjusted returns
- BAB captures this anomaly while maintaining market neutrality

#### Expected outcomes:

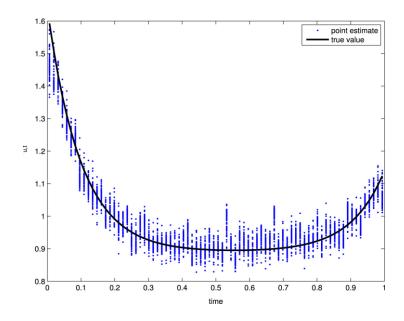
- Consistent alpha with low correlation to traditional risk factors
- Lower drawdowns in bear markets due to defensive low-beta tilt
- Enhanced Sharpe ratios when combined with volatility-based scaling



# EXPLOITING PREDICTABLE INTRADAY PATTERNS WITH 0DTE OPTIONS

## Idea

- SPX returns show predictable intraday volatility structures (diurnal patterns)
- These patterns also align with higher probabilities of jumps



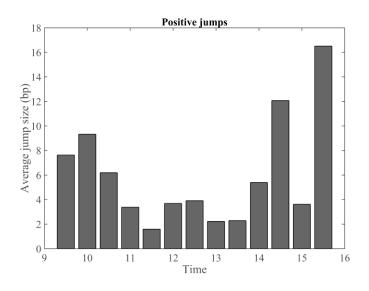


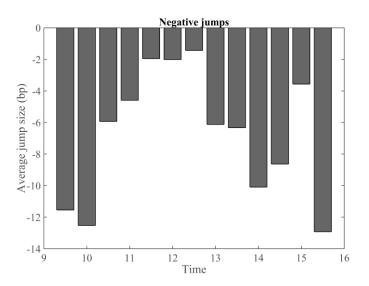
## Objective

- Use predictable intraday jump and volatility patterns to exploit at-the-money SPX 0DTE straddles
- Test whether straddle returns are positive conditional on realized volatility exceeding expected value, based on historical intraday regimes
- Develop a systematic trading model



## Jump patterns







## Strategy outline

#### Signal construction:

- Estimate in-sample U-shaped diurnal component of SPX
- Use de-seasoned return residuals to obtain in-sample jump patterns
- Estimate out-of-sample realized volatility and conditional jump probability

#### Portfolio implementation:

- Buy 1 ATM straddle when realized volatility or jump risk is elevated
- Exit based on:
  - Profit/loss threshold
  - Time stop near market close



## Data

- Intraday SPX data
- Prices of 0DTE calls and puts



## Performance expectations

#### • Why it works?

- Intraday volatility and jumps are statistically predictable
- ATM straddles profit from timing mispricings, not just magnitude
- Strategy selectively enters high-move probability windows

#### Expected outcomes:

- High return with controlled downside
- Increased Sharpe ratio during volatile, macro-sensitive regimes
- Reduced drawdown

