

Portfolio Visualizer

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Agenda

- Background and Motivation
- Portfolio Visualizer Tools for Investors
 - Overview of tools and related theoretical background
 - Investment process alignment
 - Live examples
- Limitations of Financial Modeling
- Q & A

Background and Motivation

- Empower investors and advisors with better tools
- Easy-to-use web based platform
- Focus on quantitative, factor based investing
- No product promotions or bias
- Self-guided education

Portfolio Visualizer Tools

- Portfolio and asset allocation backtesting
 - Explore portfolio construction and diversification benefits
 - Explore risk metrics and risk tolerance
- Monte Carlo simulation
 - Explore long term expected portfolio growth and portfolio survival during retirement withdrawals
- Modern portfolio theory (MPT)
 - Efficient frontier and mean variance optimization
 - Black-Litterman model
- Factor model regression analysis (CAPM, FF₃, CH₄, FF₅)
 - Understanding risk factors and fund performance attribution
- Tactical asset allocation models
 - Managing portfolio risk

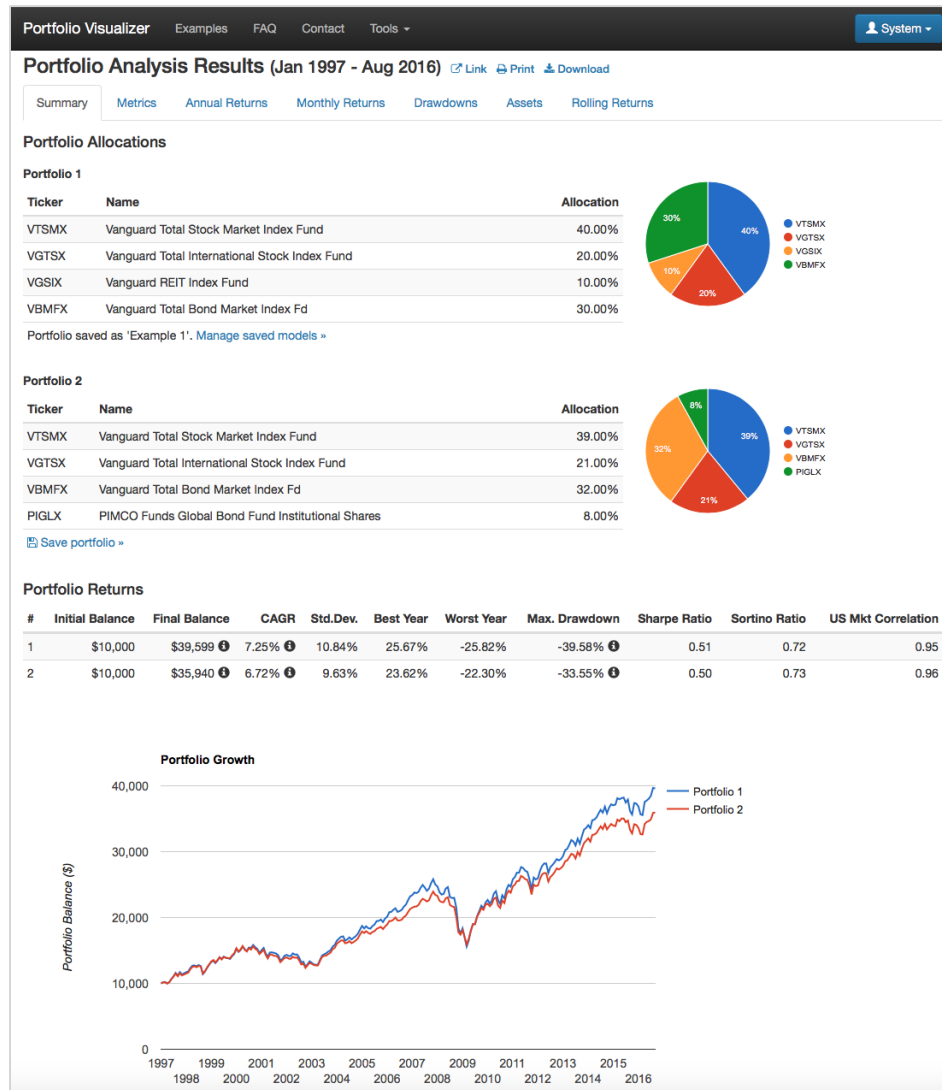
Portfolio Backtesting

- Explore return and risk characteristics of different asset allocations
 - Asset allocation is the primary driver of portfolio returns
 - Identify diversification benefits
- Easy comparisons
 - Portfolios and asset allocations
 - Rebalancing strategies
 - Common lazy portfolios
 - Nominal vs. real returns
 - Portfolio asset correlations
 - Portfolio yield and income

Risk Measures and Risk Tolerance

- Risk related measures
 - Standard deviation
 - Beta
 - Sharpe Ratio
 - Sortino Ratio
 - Treynor Ratio
 - Skewness and kurtosis of return distribution
 - Value-at-Risk (VaR)
- Personal comfort level
 - Volatility (standard deviation)
 - Worst year
 - Maximum drawdown and recovery time
 - Rolling returns
 - Positive vs. negative periods & gain/loss ratio

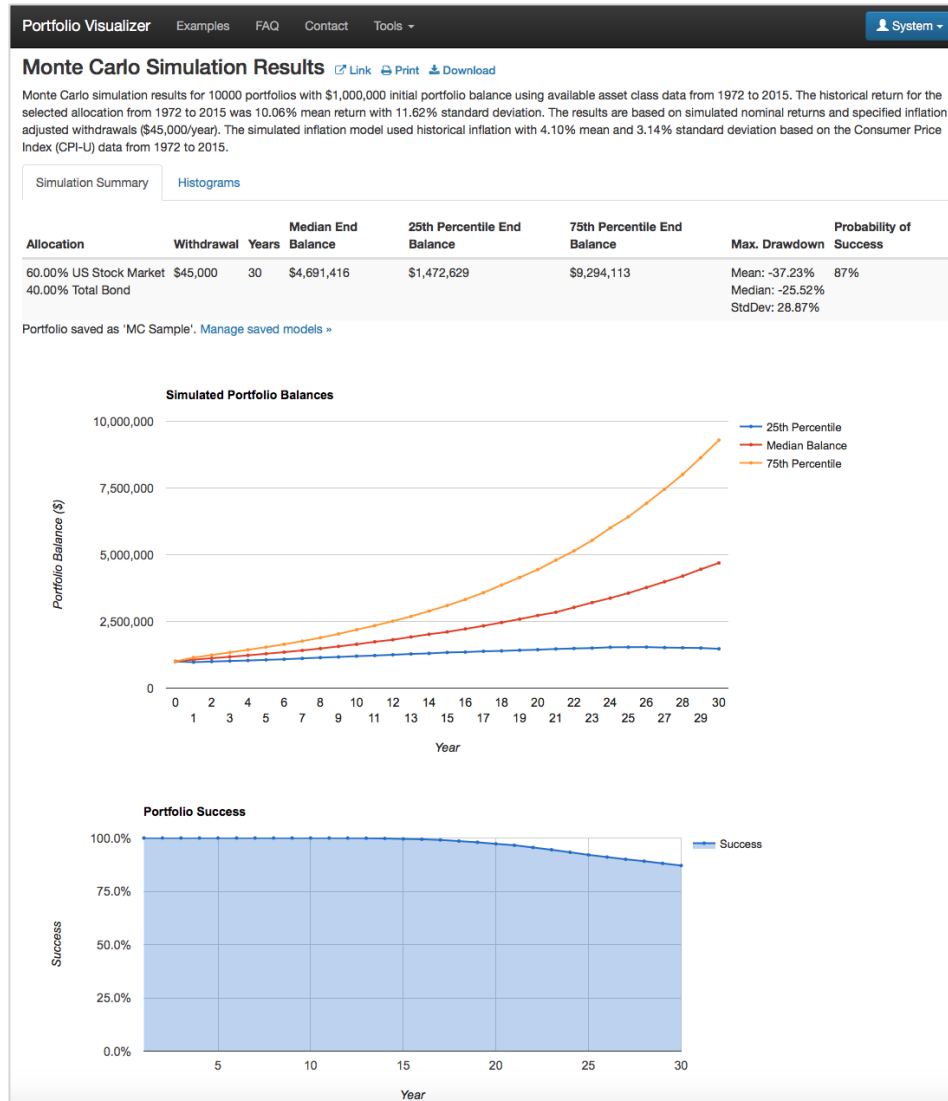
Portfolio Backtesting Example



Monte Carlo Simulation

- Explore expected future portfolio growth
- Explore sustainable withdrawals
 - Sequence risk of returns
 - Variability of outcomes
- Withdrawal models
 - Fixed withdrawals (inflation adjusted)
 - Percentage based withdrawals
 - Life expectancy based withdrawals (RMD)
- Multiple simulation models
 - Historical returns
 - Statistical distributions
 - Forecasted returns
- What-if scenario analysis
 - Forecasted inflation and asset returns

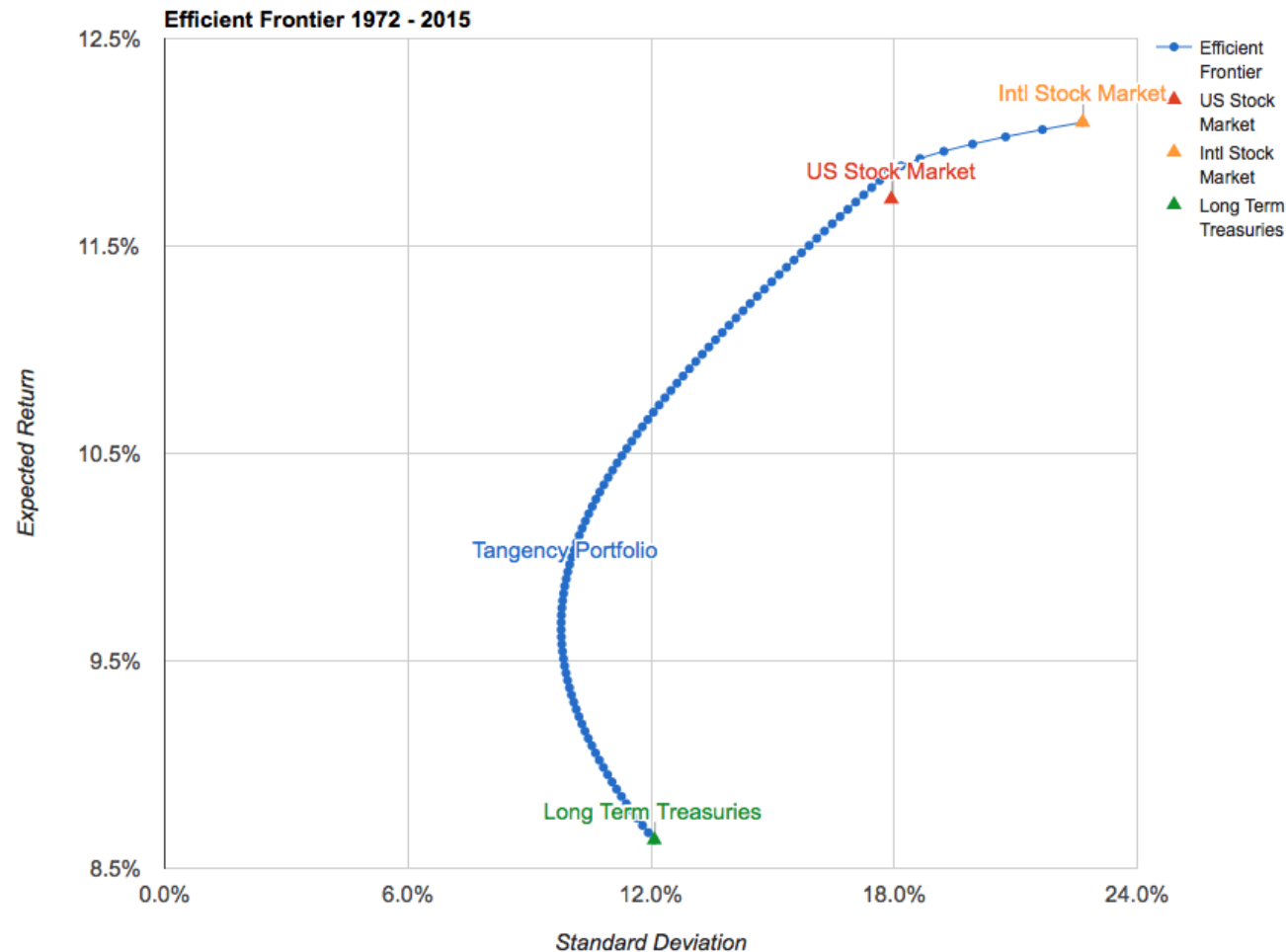
Monte Carlo Simulation Example



Modern Portfolio Theory

- Theory pioneered by Harry Markowitz on how risk-averse investors construct portfolios
 - Maximize expected return based on a given level of market risk
 - Minimize risk based on a given level of expected return
- Efficient frontier shows the portfolios that offer the maximum return for a given level of risk
 - Highlights the return and risk relationship
 - Evaluating assets in the context of overall portfolio
 - Portfolio risk level optimization

Efficient Frontier Example



Black-Litterman Model

- Main weaknesses of mean variance optimization
 - Concentrates the portfolio assets based on past performance and thus loses future diversification benefits
 - Results are typically unstable and vary significantly based on inputs (asset returns and correlations are dynamic)
- Black-Litterman model
 - No need to estimate expected asset returns
 - Derive equilibrium returns for benchmark portfolio
 - Update benchmark portfolio weights based on investor's views
 - Supports both relative and absolute views on asset returns
 - Supports confidence levels for views

Black-Litterman Model Example

[Portfolio Visualizer](#) [Examples](#) [FAQ](#) [Contact](#) [Tools](#) System

Black-Litterman Asset Allocation Model

This online portfolio optimizer tool implements the Black-Litterman asset allocation model. The Black-Litterman asset allocation model combines ideas from the Capital Asset Pricing Model (CAPM) and the Markowitz's mean-variance optimization model to provide a method to calculate the optimal portfolio weights based on the given inputs. The model first calculates the implied market equilibrium returns based on the given benchmark asset allocation weights, and then allows the investor to adjust these expected returns based on the investor's views. The opinion adjusted returns are then passed to the [mean variance optimizer](#) to derive the optimal asset allocation weights.

Step 3/3: Optimization Results

Benchmark Portfolio with Equilibrium Excess Returns

Ticker	Name	Equilibrium Return	Allocation
VTSMX	Vanguard Total Stock Market Index Fund (VTSMX)	10.86%	39.00%
VGTSX	Vanguard Total International Stock Index Fund (VGTSX)	11.94%	21.00%
VBMFX	Vanguard Total Bond Market Index Fd (VBMFX)	0.32%	32.00%
PIGLX	PIMCO Funds Global Bond Fund Institutional Shares (PIGLX)	1.90%	8.00%

Equilibrium returns are based on the expected annual return of 7.00%. Covariance matrix is based on monthly asset returns from Jun 1996 to Aug 2016.

Asset 1	View	Value	Confidence
Vanguard Total International Stock I... Select asset...	will outperform (->) ... will return	Vanguard Total Stock Market Index ... 0.00	75% 75%
Select asset...	will return	0.00	75%

[+ Add View](#)

Optimized Portfolio

Optimization Type Constrained

Ticker	Name	Adjusted Return	Allocation
VTSMX	Vanguard Total Stock Market Index Fund (VTSMX)	10.87%	40.01%
VGTSX	Vanguard Total International Stock Index Fund (VGTSX)	11.89%	19.99%
VBMFX	Vanguard Total Bond Market Index Fd (VBMFX)	0.32%	32.00%
PIGLX	PIMCO Funds Global Bond Fund Institutional Shares (PIGLX)	1.88%	8.00%

Adjusted returns are equilibrium returns adjusted for the given views. The optimized portfolio has expected return of 6.98% with annualized standard deviation of 9.86% and total allocation of 100.00%.

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Factor Models

- Explaining returns by risk factor exposures
- Supports factor regression across multiple factor models
 - Capital Asset Pricing Model (beta)
 - Fama-French 3-factor model (MKT, SMB, HML)
 - Carhart 4-factor model (MKT, SMB, HML, MOM)
 - Fama-French 5-factor model (MKT, SMB, HML, RMW, CMA)
 - Fixed income factor models (TRM, CDT)
 - Other equity factors: LT/ST Rev, QMJ, BAB, ...
 - Any combination of above factors
- Applications
 - Identifying what factors drove fund returns
 - Active fund analysis (alpha, closet index, ...)
 - Portfolio tilts to size/value

Factor Regression Example

Portfolio Visualizer

Examples

FAQ

Contact

Tools

System

Fama-French Factor Regression Analysis

This online Fama-French factor regression analysis tool supports regression analysis for individual assets or a portfolio of assets using the [capital asset pricing model \(CAPM\)](#), [Fama-French three-factor model](#), the Carhart four-factor model, or the new [Fama-French five-factor model](#). You can also run market model regression for beta analysis based on selected assets or imported benchmarks. The analysis is based on asset returns for the entered mutual funds and ETFs, and the factor returns published on [Kenneth French's web site](#) and [AQR's web site](#). The multiple linear regression indicates how well the returns of the given assets or a portfolio are explained by the Fama-French three-factor model based on market, size and value loading factors. Carhart four-factor model adds momentum as the fourth factor for explaining asset returns, and the Fama-French five-factor model extends the three-factor model with profitability (RMW) and investment (CMA) factors. The tool also supports the use of other factor models including Quality Minus Junk (QMJ) and Bet Against Beta (BAB) factors as described in [Asness-Frazzini-Pedersen papers](#). For bond funds and balanced funds you can include the fixed income factor model to explain returns based on term risk (interest rate risk) and credit risk exposure. The fixed income factors can be further adjusted to account for the yield curve and to add high yield credit risk as an additional factor. You can also view the [table of mutual fund and ETF factor regressions](#).

Regression Type ⓘ

Individual assets

Tickers ⓘ

IJS VBR RZV

Start Date ⓘ

MM/DD/YYYY (optional)

End Date ⓘ

MM/DD/YYYY (optional)

Factor Returns ⓘ

Fama-French Research Factors

Stock Market ⓘ

United States

Equity Factor Model ⓘ

Three-Factor Model

Fixed Income Factor Model ⓘ

None

Regression Basis ⓘ

Monthly Returns

Roll Period ⓘ

36 Months

Factor Analysis

Cancel

Factor Analysis Results

Factor Analysis

Residuals

Rolling Regression

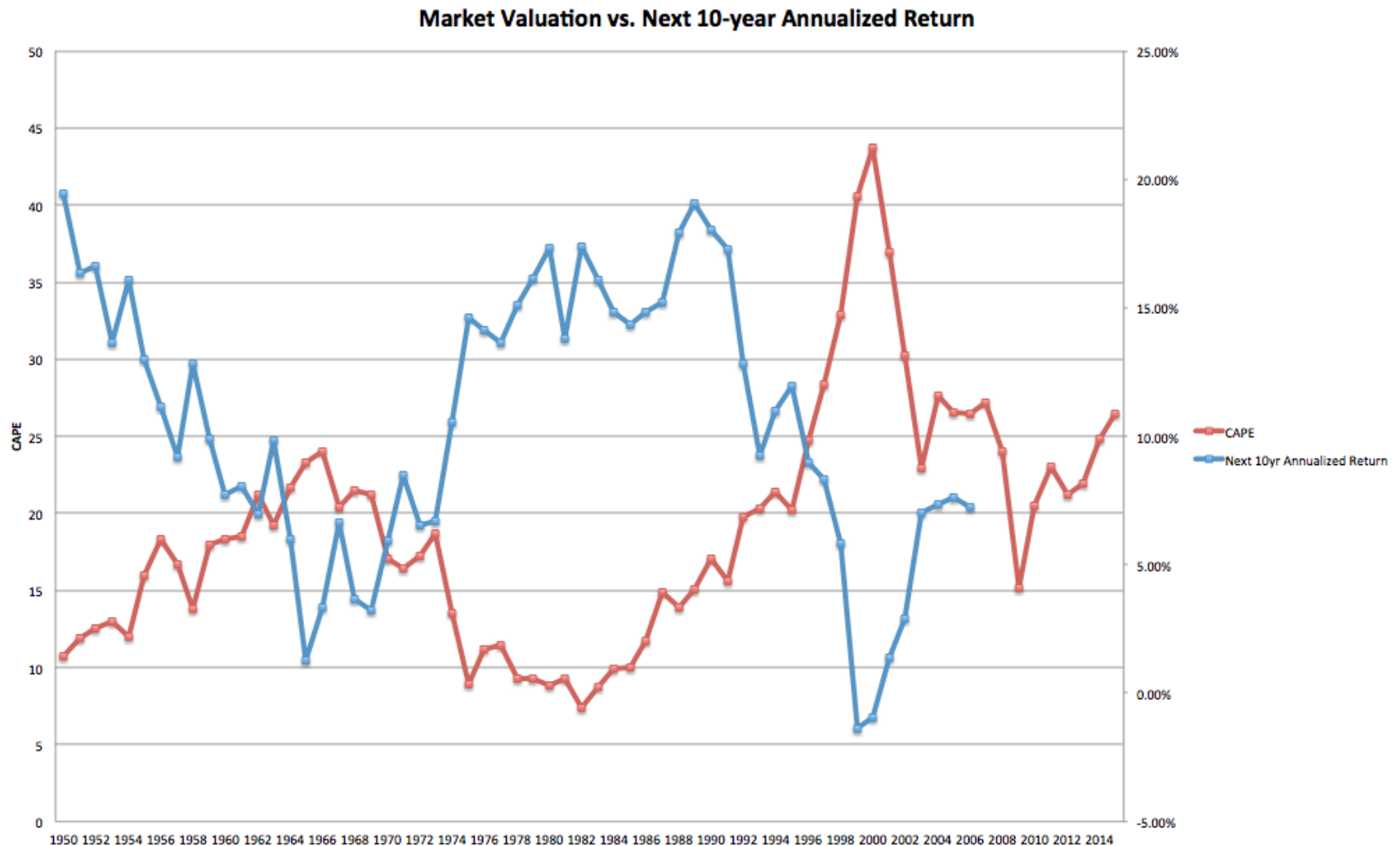
Factor Analysis Summary

Name	Ticker	Start Date	End Date	Market Exposure (B_{mkt})	Size Exposure (B_{smb})	Value Exposure (B_{hml})	Alpha (α)	Annual Alpha	R^2
iShares S&P SmallCap 600 Value ETF	IJS	Aug 2000	Jun 2016	0.95	0.81	0.47	0.06%	0.68%	95.3%
Vanguard Small-Cap Value ETF	VBR	Feb 2004	Jun 2016	1.03	0.60	0.38	0.06%	0.68%	97.2%
Guggenheim S&P Smallcap 600 Pure Value ETF	RZV	Apr 2006	Jun 2016	1.20	1.25	1.09	0.03%	0.37%	84.0%

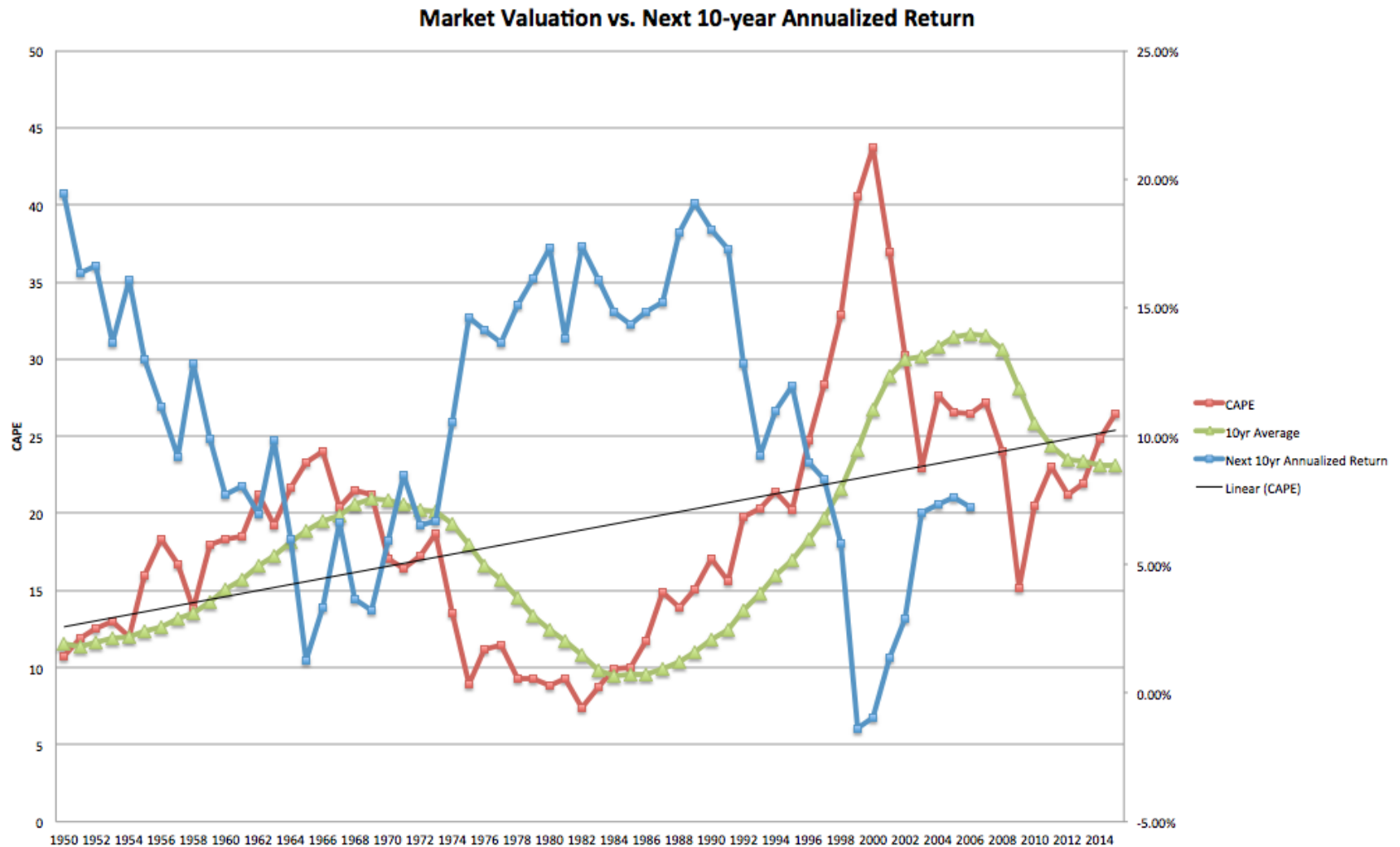
Tactical Asset Allocation

- Explore tactical asset allocation models that aim to provide better risk adjusted returns
 - Protecting capital during major drawdown events
- Models can be based on multiple techniques such as
 - Economic and fundamental indicators
 - Technical indicators
 - Sentiment indicators
 - Volatility indicators
 - Combinations of above
- Many popular technical indicators are based on momentum
 - Premier market anomaly observed across multiple asset classes and market regions
- Tactical asset allocation can be controversial
 - Market timing is seldom easy and reliable
 - Many published models are subject to data mining and over optimization
 - Many models use binary on/off risk model that assumes 100% confidence in signals
 - Tactical asset allocation models can have long periods of underperformance
 - Tax implications can have big impact on tactical asset allocation model returns
 - Multiple research papers both in favor and against market timing

Valuation Model Example



Valuation Model Example



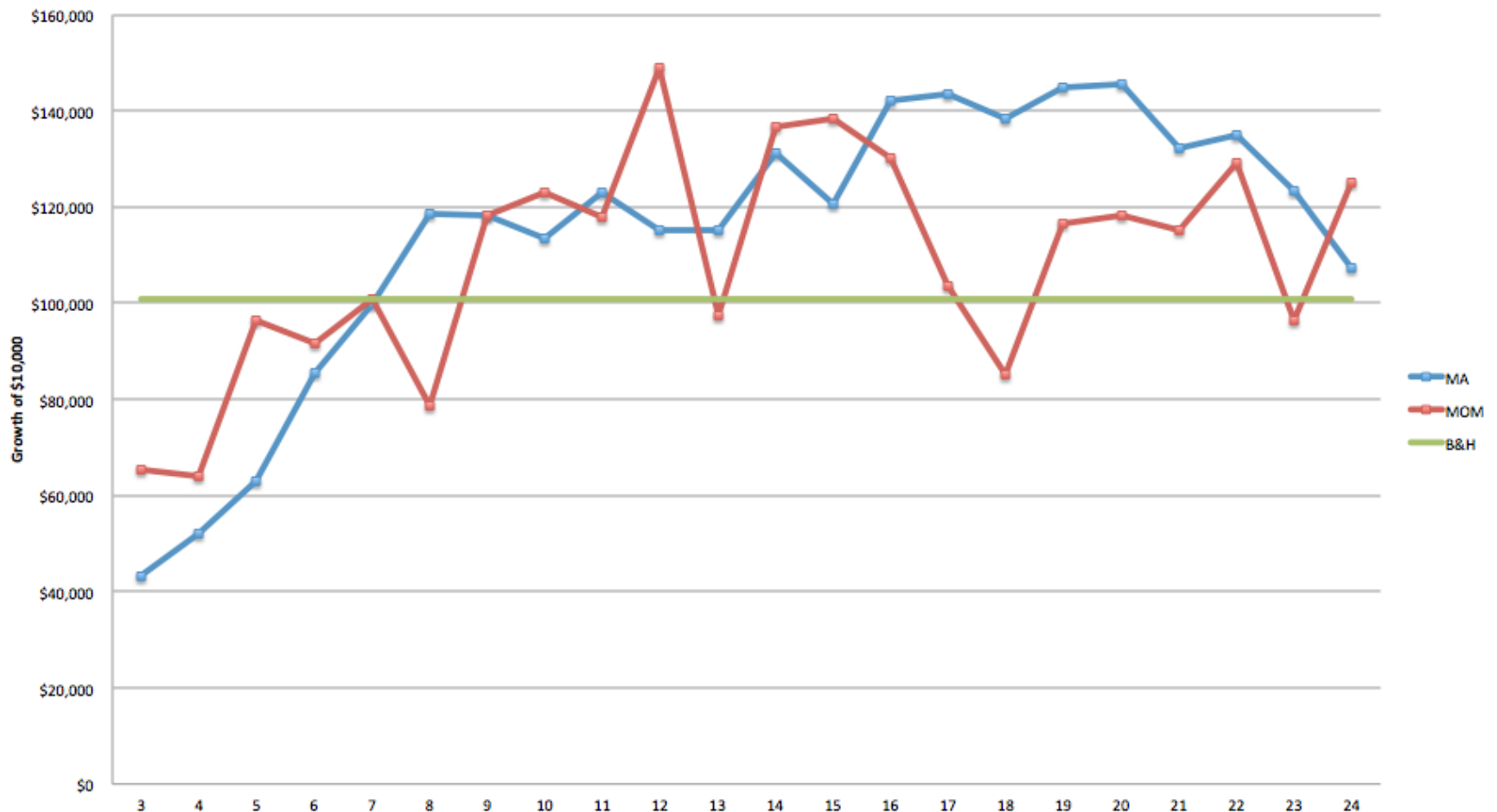
Momentum Examples

	Domestic Stocks			International Stocks			Real Estate			Commodities			Bonds		
	B&H	MA	MOM	B&H	MA	MOM	B&H	MA	MOM	B&H	MA	MOM	B&H	MA	MOM
CAGR	11.74%	11.56%	11.78%	9.44%	10.62%	9.69%	13.67%	13.85%	14.43%	5.64%	7.88%	8.63%	8.55%	8.43%	9.17%
StDev	15.01%	12.00%	11.90%	17.04%	12.45%	11.86%	16.92%	12.17%	12.19%	19.15%	15.19%	15.23%	8.37%	7.11%	5.94%
Sharpe	0.50	0.57	0.59	0.33	0.49	0.44	0.56	0.74	0.78	0.13	0.26	0.31	0.45	0.50	0.70
Max DD	-50.21%	-23.58%	-29.58%	-56.68%	-21.07%	-25.72%	-68.30%	-20.78%	-19.98%	-69.38%	-52.38%	-55.02%	-20.97%	-11.26%	-6.41%

- 12 month moving average and time series momentum comparison across asset classes from January 1976 to December 2014
 - Used with permission from DIY Financial Advisor by David Foulke, Jack Vogel, and Wesley Gray (Wiley, 2015)

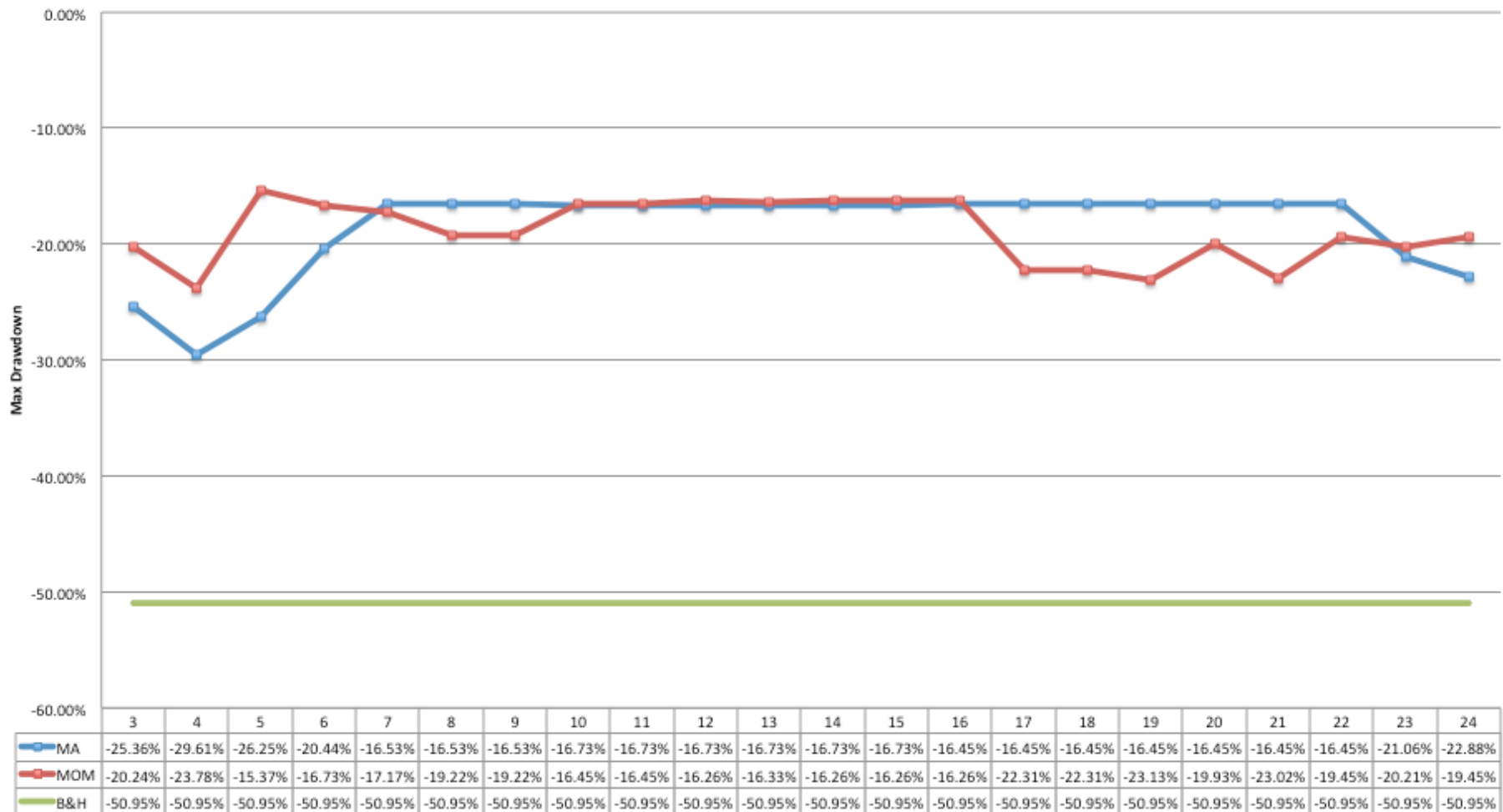
Momentum Examples

S&P 500 TR Growth of \$10,000 by Timing Period 1990 - 2015

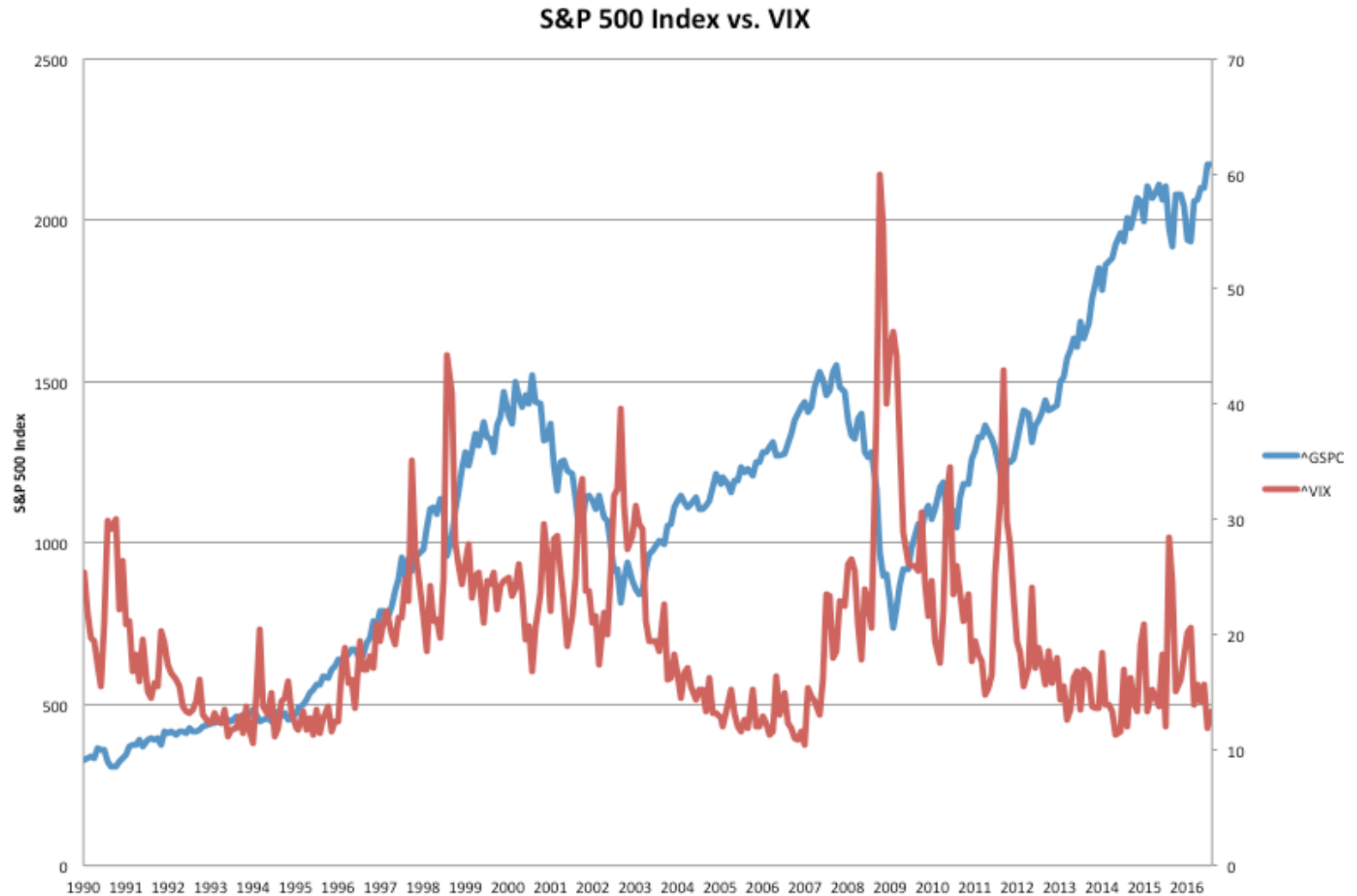


Momentum Examples

S&P 500 TR Max Drawdown by Timing Period 1990 - 2015

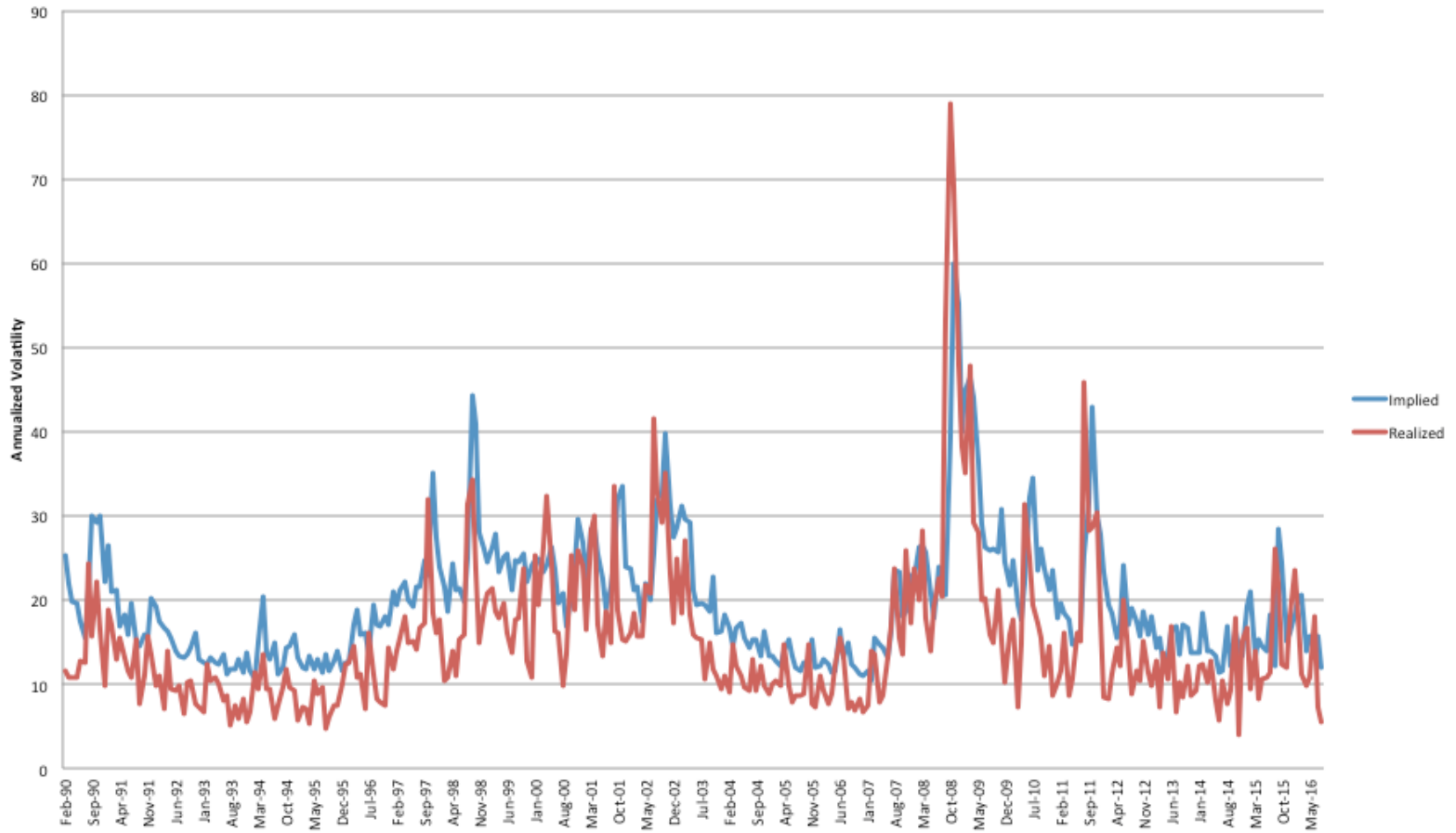


Volatility Example

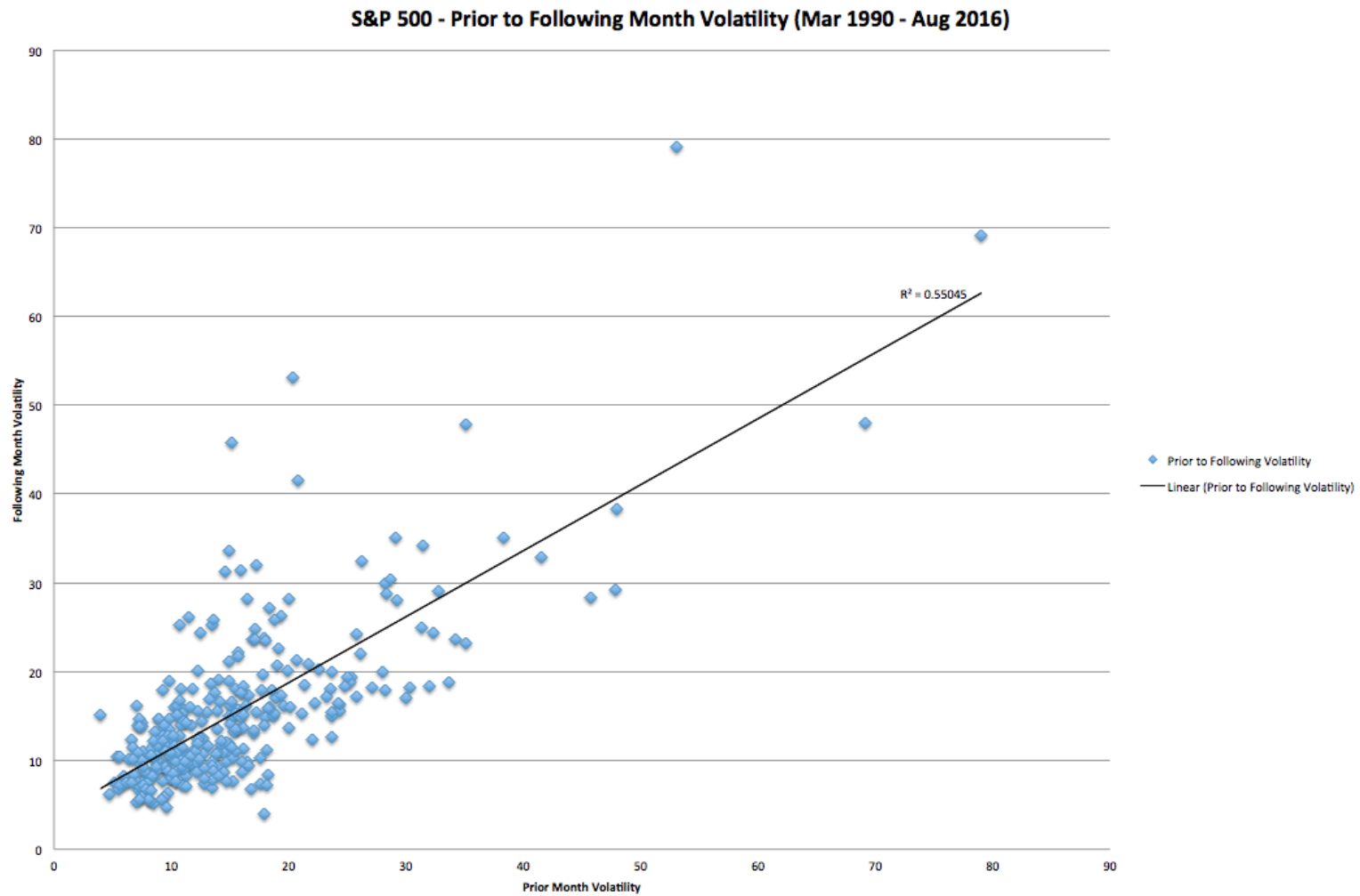


Volatility Example

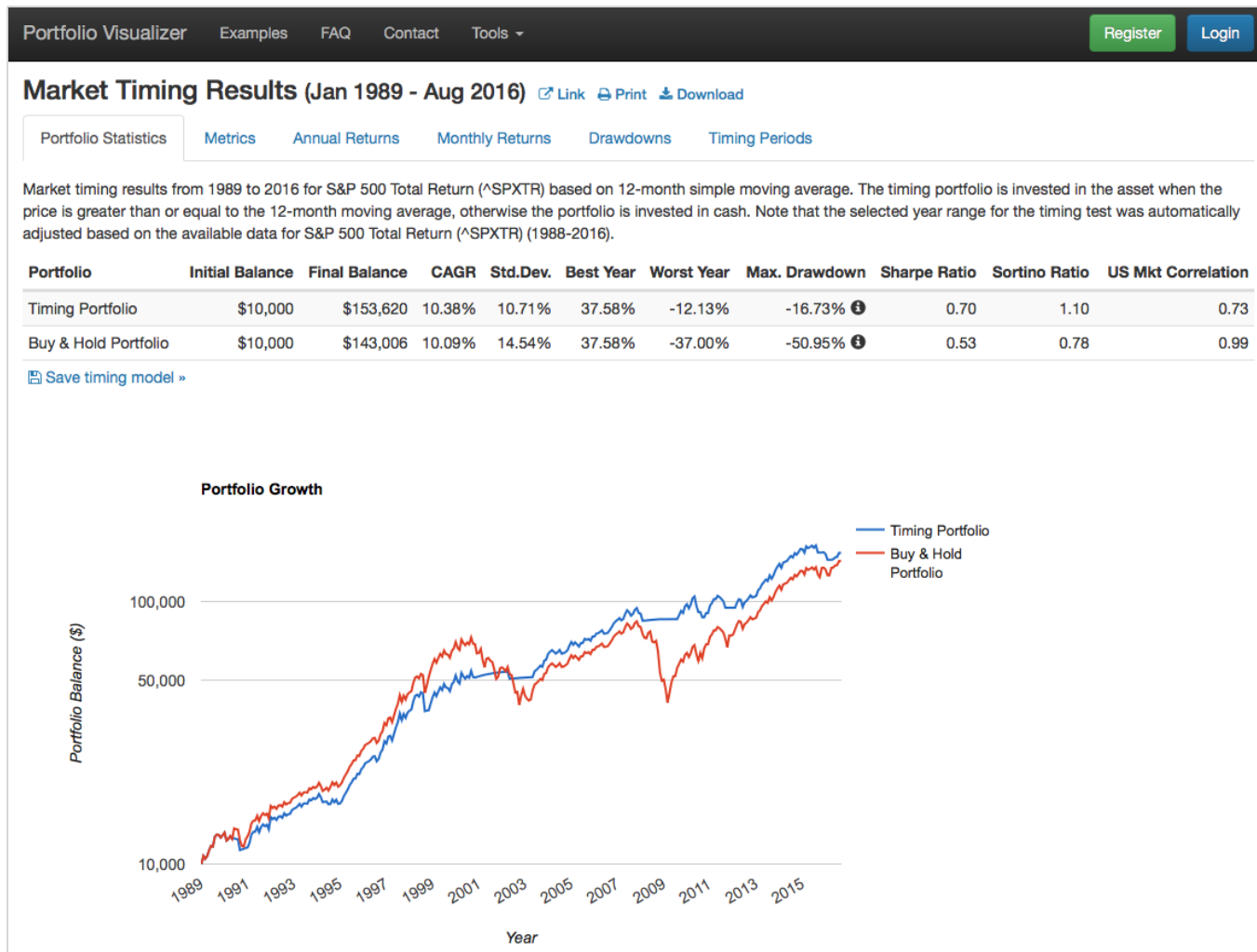
S&P 500 - Implied vs. Realized Volatility



Volatility Example



Tactical Asset Allocation Examples



Financial Modeling Limitations

- Models and related tools can be helpful in understanding concepts and market dynamics
- Understanding limitations and assumptions behind models is important
- Common limitations and issues
 - Assumptions on return distribution (skewness, kurtosis)
 - Limited amount of historical data for statistical analysis
 - Asset returns and correlations are dynamic and change over time
 - Fundamental changes in the macro environment
 - Interest rates
 - Future returns may be different than past returns
 - Unexpected events
 - Taxes and trading costs are typically not reflected in the results
 - Model may suffer from biases
 - Data mining and data fitting bias
 - Sample selection bias
 - Survivorship bias
 - Look ahead bias

Questions and Answers

