Portfolio Visualizer

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- 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, advisors and investment managers with better tools
- Focus on quantitative, factor based investing
- Easy-to-use web based platform
- No product promotions or bias
- Support 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
- Portfolio optimization
 - Modern portfolio theory
 - Efficient frontier and mean variance optimization
 - Black-Litterman model
 - Risk reduction (CVaR, risk parity)
- 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
 - Nominal vs. real returns
 - Portfolio asset correlations
 - Portfolio yield and income
- Impact of contributions and withdrawals
- Risk and return decompositions
- Portfolio exposures
 - Equity break-downs by market capitalization, industry sector, and region
 - Fixed income break-downs by credit quality and maturity
- Impact of management fees

Portfolio Risk And Return

Risk and return measures

- CAGR, standard deviation
- Beta, Market correlation
- Sharpe, Sortino and Treynor Ratios
- Skewness and kurtosis of return distribution
- Value-at-Risk (VaR), Conditional Value-at-Risk (CVaR)
- Drawdowns and recovery times
- Risk and return contributions of portfolio assets
- Rolling returns
- Worst year performance
- Positive vs. negative periods & gain/loss ratio
- Statistical metrics vs. personal risk tolerance

Portfolio Backtesting Example

Portfolio	Visualizei	Examples	FAQ	Contact	Tools -					L System -
Portfol	io Ana	lysis Resi	u lts (Jan	1997 -	Aug 2016) 🕑 Link 🔒 F	Print 🛓 Download			
Summary	/ Metric	s Annual R	eturns M	Monthly Ret	urns Dra	wdowns A	ssets Rolling Ret	urns		
Portfolio	Allocatio	ons								
Portfolio 1										
Ticker	Name						Allocation			
VTSMX	Vangua	rd Total Stock M	arket Index F	und			40.00%	30%	40% 0 VTS	
VGTSX	Vangua	rd Total Internatio	onal Stock In	dex Fund			20.00%		● VGS ● VBM	
VGSIX	Vangua	rd REIT Index Fu	nd				10.00%	20%		
VBMFX	Vangua	rd Total Bond Ma	arket Index F	d			30.00%			
Portfolio 2		ample 1'. Manag	e saved mod	leis »				8%		
Ticker	Name						Allocation		• vts	мх
VTSMX VGTSX	-	I Total Stock Mai					39.00%	32%	39% • VGT • VGT	
VBMFX	*	Total Bond Mar					32.00%		• PiGL	
PIGLX		unds Global Bor			700		8.00%	21%		
🖺 Save por	rtfolio »									
Portfolio		Final Balance	CAGR	Std.Dev.	Best Year	Worst Year	Max. Drawdown	Sharpe Ratio	Sortino Ratio	US Mkt Correlation
Portfolio	Returns	Final Balance \$39,599 3	CAGR 7.25% 3	Std.Dev.	Best Year 25.67%	Worst Year -25.82%		Sharpe Ratio 0.51	Sortino Ratio	US Mkt Correlation 0.95
Portfolio # Initial	Returns Balance						Max. Drawdown			
Portfolio # Initial 1	Returns Balance \$10,000	\$39,599 6 \$35,940 6 Portfolio Gro	7.25% 3 6.72% 3	10.84%	25.67%	-25.82% -22.30%	Max. Drawdown -39.58% 🕄 -33.55% 🕄	0.51	0.72 0.73	0.95
Portfolio # Initial 1 2	Returns Balance \$10,000 \$10,000	\$39,599 () \$35,940 () Portfolio Gro	7.25% 3 6.72% 3	10.84%	25.67%	-25.82% -22.30%	Max. Drawdown -39.58% 🕄 -33.55% 🕄	0.51 0.50	0.72 0.73	0.95
Portfolio # Initial 1	Returns Balance \$10,000 \$10,000 30,000 20,000	\$39,599 @ \$35,940 @ Portfolio Gro	7.25% 3 6.72% 3	10.84%	25.67%	-25.82% -22.30%	Max. Drawdown -39.58% 🕄	0.51 0.50	0.72 0.73	0.95
Portfolio # Initial 1	Returns Balance \$10,000 \$10,000 30,000 20,000 10,000	\$39,599 \$35,940 Portfolio Gro	7.25% ⁽²⁾ 6.72% ⁽²⁾ wth	10.84%	25.67% 23.62%	-25.82% -22.30%	Max. Drawdown -39,58% • -33,55% •	0.51 0.50	0.72 0.73	0.95

Monte Carlo Simulation

- Explore expected future portfolio growth and 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
- Retirement planning
 - Portfolio glide paths
 - Multiple time dependent cashflows
 - Stress test sequence of returns risk

Monte Carlo Simulation Example

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Monte Carlo Simulation Results CLink Deprint & Download

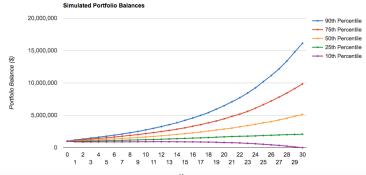
Monte Carlo simulation results for 10000 portfolios with \$1,000,000 initial portfolio balance using available historical returms data from Jan 1972 to Dec 2017. The historical return for the selected portfolio for this period was 10.21% mean return (9.64% CAGR) with 9.93% standard deviation of annual returns. The simulation results are based on generated nominal returns and specified inflation adjusted withdrawais (545,000 per year). The simulated inflation model used historical inflation with 3.98% mean and 1.32% standard deviation based on the Consumer Price Index (CPI-U) data from Jan 1972 to Dec 2017. The generated inflation samples were correlated with simulated asset returns based on historical correlations. The simulation time period was constrained by the available history for US Stock Market (Jan 1972 - Dec 2017). © Save simulation model »



Summary Statistics

	10th Percentile	25th Percentile	50th Percentile	75th Percentile	90th Percentile
Portfolio End Balance (nominal)	\$0.00	\$2,079,274	\$5,129,083	\$9,866,921	\$16,142,576
Portfolio End Balance (inflation adjusted)	\$0.00	\$626,910	\$1,608,965	\$3,211,340	\$5,308,130
Maximum Drawdown	-100.00%	-38.45%	-27.84%	-22.13%	-19.36%
Maximum Drawdown Excluding Cashflows	-30.90%	-25.34%	-21.82%	-19.36%	-16.60%

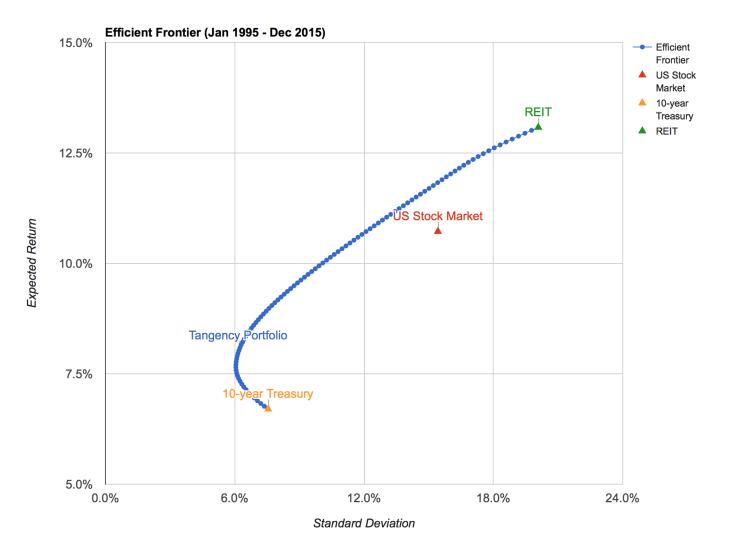
8985 portfolios out of 10000 simulated portfolios (89.85%) survived all withdrawals.



Portfolio Optimization

- Modern portfolio theory by Harry Markowitz
 - Maximize expected return based on portfolio risk
 - Minimize portfolio risk based on expected return
- Efficient Frontier plots with historical or forecasted returns
 - Highlights the return and risk relationship
 - Evaluating assets in the context of overall portfolio
- Unconstrained vs. constrained optimization
- Rolling optimization
- Other optimization strategies:
 - Expected tail loss (CVaR), risk budgeting, target volatility

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)
- Ignores skewness and kurtosis of asset returns
- 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

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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 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%



👤 System -

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.

Investor Views and Opinion Based Adjustments

Asset 1	View		Value		Confidence	
Vanguard Total International Stock I 🔻	will outperform (->) *	Vanguard Total Stock Market Index *	1.00	%	75%	•
Select asset *	will return 🔻		0.00	%	75%	*
Select asset v	will return 🔻		0.00	%	75%	*

+ Add View

Optimized Portfolio

Optimiza	tion Type 😉	Constrained			*	
Ticker	Name			Adjusted	Return	Allocation
VTSMX	Vanguard Total Stock Market Index I	Fund (VTSMX)			10.87%	40.01%
VGTSX	Vanguard Total International Stock In	ndex Fund (VGTSX)			11.89%	19.99%
VBMFX	Vanguard Total Bond Market Index F	d (VBMFX)			0.32%	32.00%
PIGLX	PIMCO Funds Global Bond Fund Ins	stitutional Shares (PIC	GLX)		1.88%	8.00%
6.98% w	I returns are equilibrium returns adjuste rith annualized standard deviation of 9. portfolio »	-		ortfolio has ex	pected ret	urn of
		Update Views	« Previous	Restart	Cancel	

Factor Models

- Explain returns by risk factor exposures
 - Performance attribution of fund returns
 - Portfolio tilts to size/value
 - Active fund analysis (active fund manager alpha, closet index, ...)
 - Smart beta ETFs and related factor exposures
- Factor regression across multiple factor models
 - Capital Asset Pricing Model (market beta)
 - Fama-French 3-factor model (MKT-Rf, SMB, HML)
 - Carhart 4-factor model (MKT-Rf, SMB, HML, MOM)
 - Fama-French 5-factor model (MKT-Rf, SMB, HML, RMW, CMA)
 - Fixed income factor models (TRM, CDT)
 - Other equity factors (quality, low beta, ...) and custom combinations

Factor Regression Example

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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 6	IJS VBR RZV	Q	
Start Date	MM/DD/YYYY (optional)		
End Date 0	MM/DD/YYYY (optional)		
Factor Returns	Fama-French Research Factors		Ŧ
Stock Market	United States		
Equity Factor Model	Three-Factor Model		
Fixed Income Factor Model ⁽¹⁾	None •		
Regression Basis 0	Monthly Returns *		
Roll Period 3	36 Months *		
	Factor Analysis Cancel		

Factor Analysis Results CLink Deprint & Download

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

Market Valuation

Market Valuation vs. Next 10-year Annualized Return



Momentum Examples

	Don	nestic St	ocks	Interr	national S	tocks	F	Real Estat	e	Co	ommoditi	es		Bonds	
	B&H	MA	MOM	B&H	МА	MOM	B&H	MA	МОМ	B&H	MA	МОМ	B&H	MA	мом
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)

Tactical Asset Allocation Examples

Market Timii	ig Results	(Jan 1989 -	· Aug 2	016) 🗷	Link 🔒 Prin	t 📥 Downloa	d				
Portfolio Statistics	Metrics A	nnual Returns	Monthl	ly Returns	Drawdo	wns Timi	ing Periods				
Market timing results price is greater than o idjusted based on the	r equal to the 12-n	nonth moving av	erage, oth	erwise the	portfolio is ir						
Portfolio	Initial Balance	Final Balance	CAGR	Std.Dev.	Best Year	Worst Year	Max. Drawdown	Sharpe Ratio	Sortino Ratio	US Mkt Co	rrelatior
Timing Portfolio	\$10,000	\$153,620	10.38%	10.71%	37.58%	-12.13%	-16.73% 🕄	0.70	1.10		0.73
Buy & Hold Portfolio	\$10,000	\$143,006	10.09%	14.54%	37.58%	-37.00%	-50.95% 🕄	0.53	0.78		0.99
	Portfolio Grov	vth						 Timing Portfoli Buy & Hold Portfolio 	o		
100,00 છ		vth							o		
100,00 bortfolio Balance (\$) 20,00	0	wth	Ym	by for		WWW WWW		Buy & Hold	o		

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, volatilities and correlations are dynamic and change over time
 - Fundamental changes in the macro environment (e.g. 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

