

REGIME AWARE ASSET ALLOCATION AND ASSET-LIABILITY MANAGEMENT



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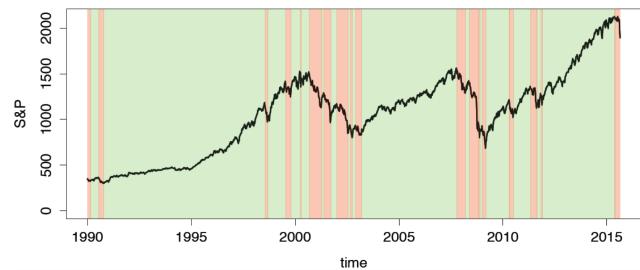
PRINCETON UNIVERSITY

APRIL 12, 2018

Agenda

1. Factor investing for asset allocation and ALM
 - Motivation
 - Apply machine learning concepts
2. Regime aware factor investing and ALM
 - Motivate economic regimes
 - Factors within regimes
3. Comparing Pension Systems
 - Importance of realistic goals

Two-Regimes for the S&P 500 Index (1990-2015)



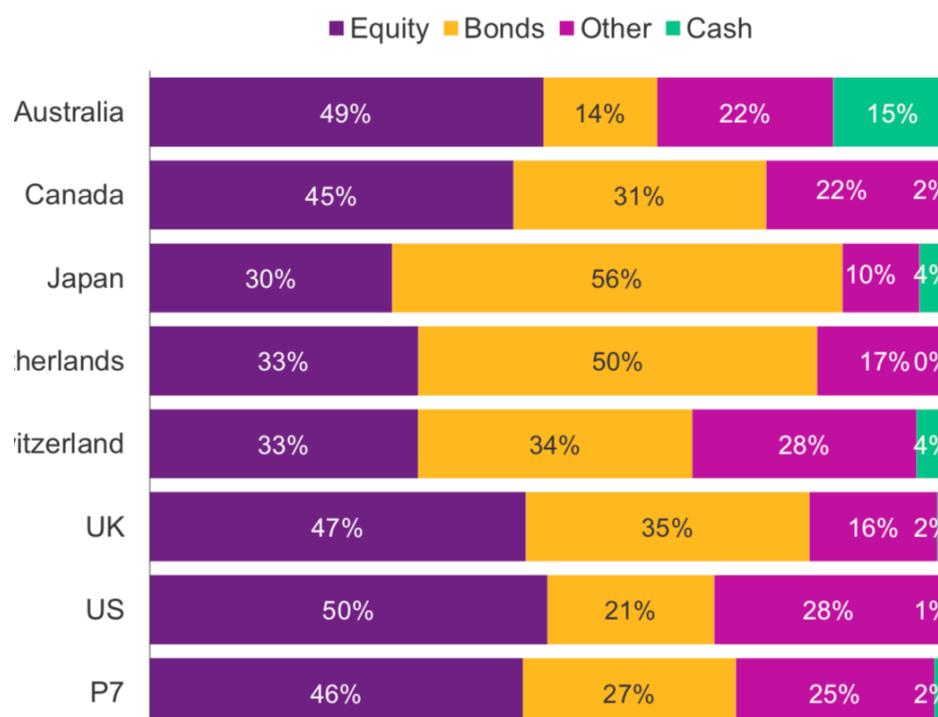
Pension Systems throughout the World

Market	Total Assets 2017 (USD billion)	Assets/GDP ratio (%) ⁷
Australia	1,924	138.4%
Brazil ¹	269	12.9%
Canada	1,769	107.8%
Chile	205	77.8%
China ²	177	1.5%
Finland	233	92.8%
France	167	6.5%
Germany ³	472	12.9%
Hong Kong	164	49.1%
India	120	4.9%
Ireland	157	48.2%
Italy	184	9.6%
Japan ⁴	3,054	62.5%
Malaysia	227	73.4%
Mexico	177	15.5%
Netherlands	1,598	193.8%
South Africa	258	75.1%
South Korea	725	47.4%
Spain	44	3.3%
Switzerland ⁵	906	133.1%
UK	3,111	121.3%
US ⁶	25,411	131.2%
Total	41,355	67.0%⁸

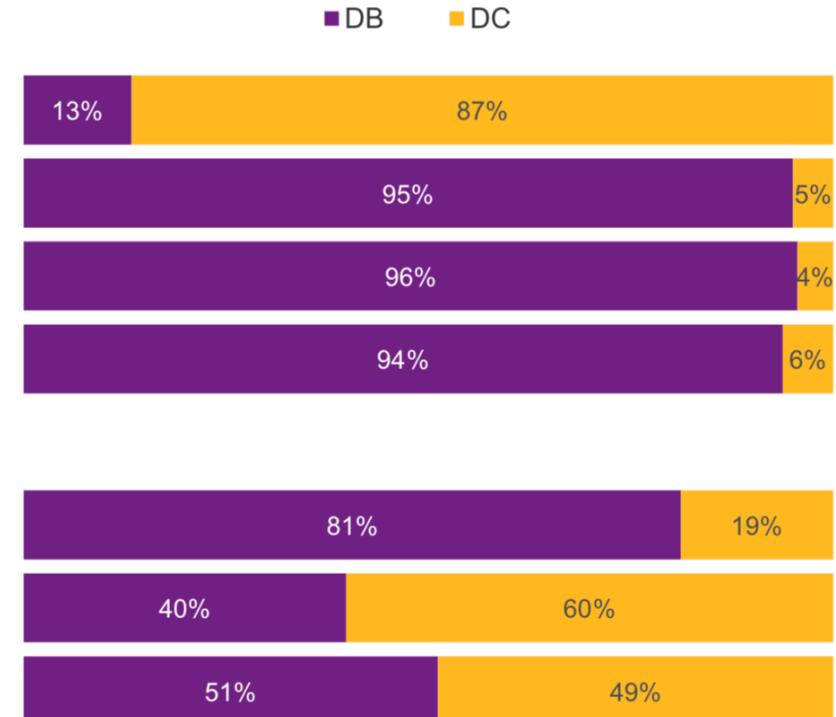
Source: Willis Towers Watson and secondary sources

Global Pension Assets Study 2017

Asset allocation 2017



DB/DC split 2017^{1,2}



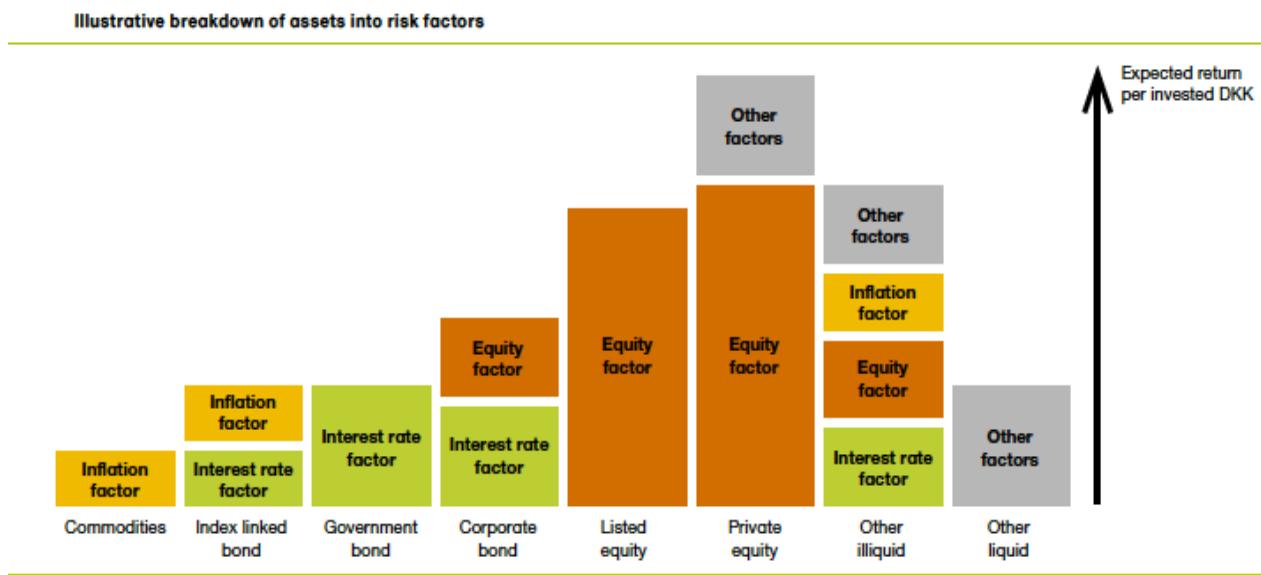
Source: Willis Towers Watson and secondary sources

Factor Investing

- ***Motivation***
 - Intensive search for higher returns to achieve goals and meet liabilities
- 1. Improve diversification -- get paid to accept (sell) risks
 - Insurance analogy
- 2. Help explain newer hybrid securities and asset categories
 - Examples:
 - High yield bonds
 - Dynamic smart beta strategies (long short via futures)
- 3. Assist during crash periods when contagion is present
 - Search for truly diversifying assets

Example: Factors as Building Blocks (critical ingredients)

Danish Pension System ATP – Factors for Assets (Ang)



Liability-related factors – Real economic growth, inflation (hard to link to asset returns)

Why Factor Investing?

- Many institutional investors have made the shift to “alternative” asset categories
 - 57% on average for U.S. endowments above \$1 B
 - Main categories: **private equity, real assets, hedge funds**
 - There is great diversity in this domain
 - The newer asset categories capture many types of risks

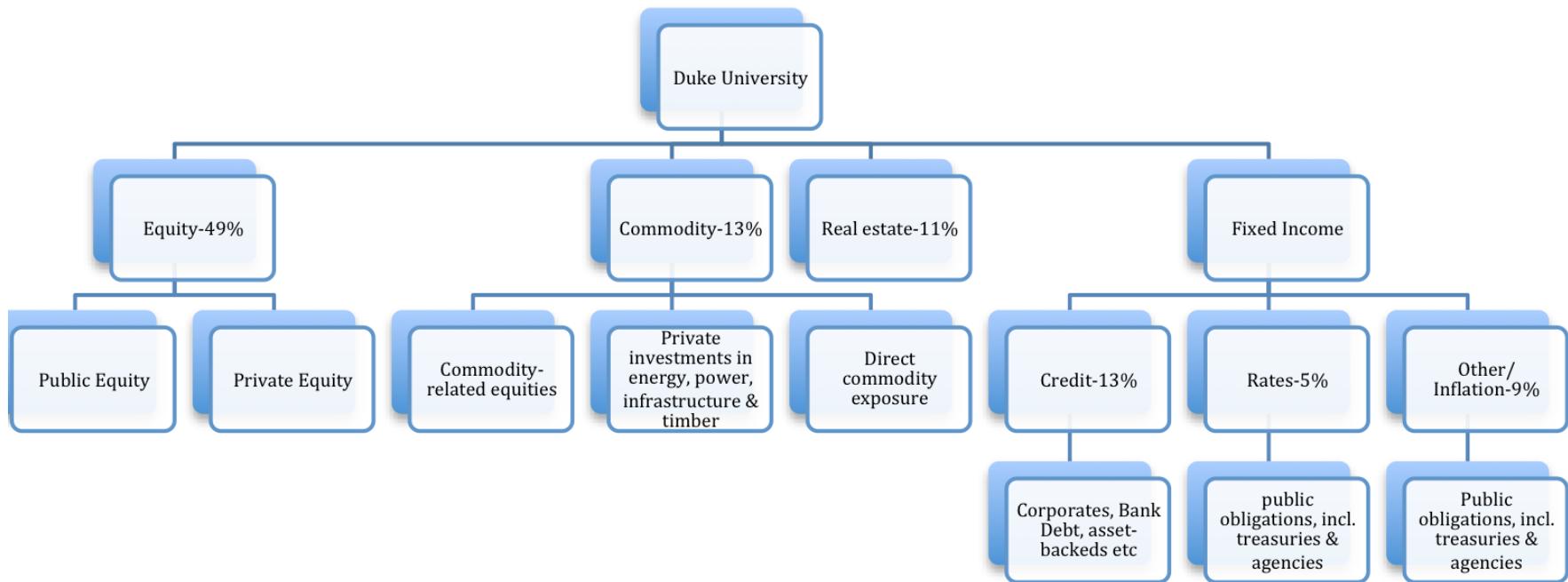
Average Allocation Across University Endowments

Asset Allocation for U.S. Colleges and Universities 2014
(NACUBO 2015)

	Survey Average	Endowments over \$1b
Equities	36%	31%
<i>Domestic Equities</i>	17%	13%
<i>International Equities</i>	19%	18%
Fixed Income	9%	8%
Alternatives	51%	57%
Short-term securities/cash/other	4%	4%

Diversity in Defining Asset Categories

Duke's Target Asset Allocation & Asset Category Descriptions (June 30 2014)
Source: Duke (2014)



Alternative Factor Approaches

(Feature selection in machine learning)

- Purely statistical factors
 - Factor analysis, principle component analysis
- Fundamental macro economic
 - Chen, Roll, and Ross
 - Maturity premium (long- short government bonds), expected inflation, unexpected inflation, industrial production growth, and default premium (corporate high versus low grade bonds)
- Micro factors
 - Fama and French
 - Equity markets risks, small minus large stock returns, value minus growth
 - Profitability (high-low profit), and investment (conservative-aggressive investment)
 - Momentum, and low volatility
- Number of factors determined by sensitivity analysis
 - Most studies have shown that about 5 factors are best (little improvement above 5 for equities)

Factor Loadings for Harvard Endowment

Exhibit 4. Example of an asset class mapping matrix

	World Equities	U.S. Treasuries	High Yield	Inflation Protection	Currency Protection
U.S. Equities	1.0			0.1	0.5
Foreign Equities	1.0			0.1	-0.5
Private Equity	1.3		0.2	0.1	0.3
Real Assets	0.3		0.8	0.3	
Commodities				2.0	-0.5
U.S. Treasuries		1.0			
TIPS		1.0		1.0	
Corporate Bonds		0.8	0.2		
Foreign Bonds		0.8			-1.0
Absolute Return	0.2		0.2		

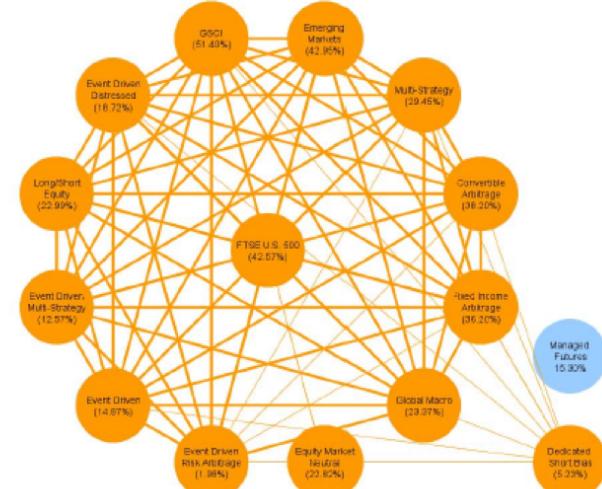
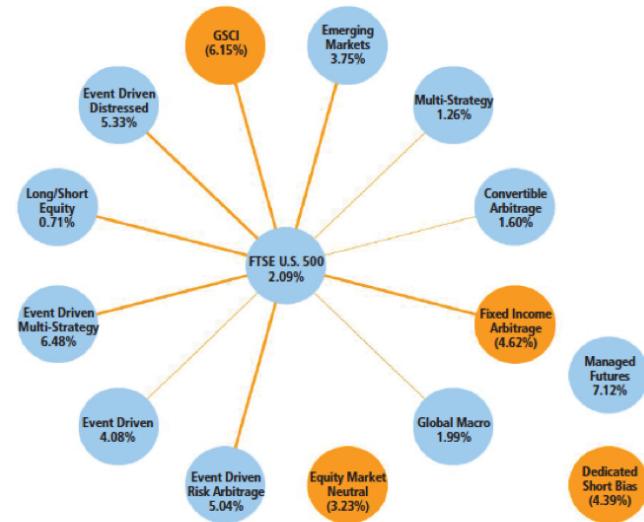
Apply cross validation from machine learning

Motivate Regimes with a Study of a University Endowment

Most Hedge Funds Experienced Contagion during 2008 Crash → massive change in covariance matrix

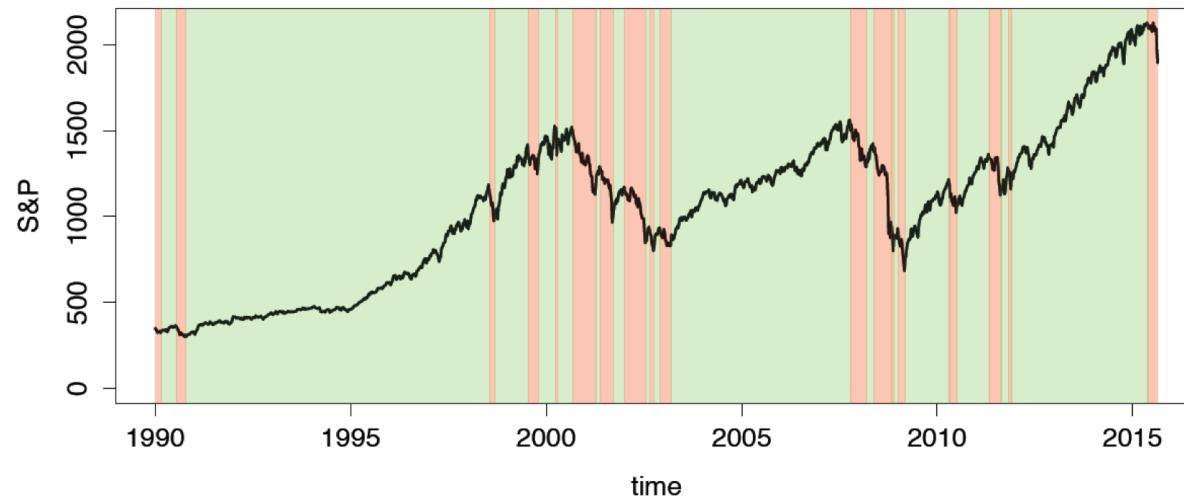
Most Hedge Funds Experienced the Classic Pattern of Contagion during the 2008 Crash
Note: stark differences between normal 2001-2007 period (left side) and crash 2008 period (right side)

(heavy line = correlation > .5; light line = correlation between .2 and .5; no line = correlation < .2)



Identify Two Regimes via Trend Filtering Algorithm*

Two-Regimes for the S&P 500 Index (1990-2015)



*See appendix

Historical Performance

	Private Equity	Real Estate	Hedge Fund	Real Assets	U.S. Equities	International Equity - Developed	International Equity - Emerging	U.S. Government Bond
Annualized Rate								
Single Regime Returns	6.50%	5.50%	5.00%	4.00%	4.50%	4.20%	4.80%	1.00%
Annual Volatility	14.212%	9.827%	7.681%	6.045%	17.988%	19.503%	26.344%	5.546%

Panel C: Historical Returns for Assets under Growth Regime
(Inflation adjusted)

	Private Equity	Real Estate	Hedge Fund	Real Assets	U.S. Equities	International Equity - Developed	International Equity - Emerging	U.S. Government Bond
Annualized Rate								
Return under growth	15.00%	9.00%	8.50%	4.50%	17.00%	16.50%	18.00%	0.30%
Annual Volatility	12.089%	7.881%	5.887%	6.157%	13.556%	15.731%	23.762%	4.236%

Panel D: Historical Returns for Assets under Contraction Regime
(Inflation adjusted)

	Private Equity	Real Estate	Hedge Fund	Real Assets	U.S. Equities	International Equity - Developed	International Equity - Emerging	U.S. Government Bond
Annualized Rate								
Return under crash	-21.66%	-7.41%	-7.91%	2.02%	-33.50%	-33.31%	-34.80%	3.85%
Annual Volatility	9.455%	12.345%	8.678%	5.561%	12.898%	13.654%	18.716%	7.601%

Transition Matrix

Equilibrium Transition Matrix
(Probability, period t to $t+1$)

	Growth Regime at Time $t+1$	Contraction Regime at Time $t+1$
Growth Regime at Time t	0.9	0.1
Contraction Regime at Time t	0.4	0.6

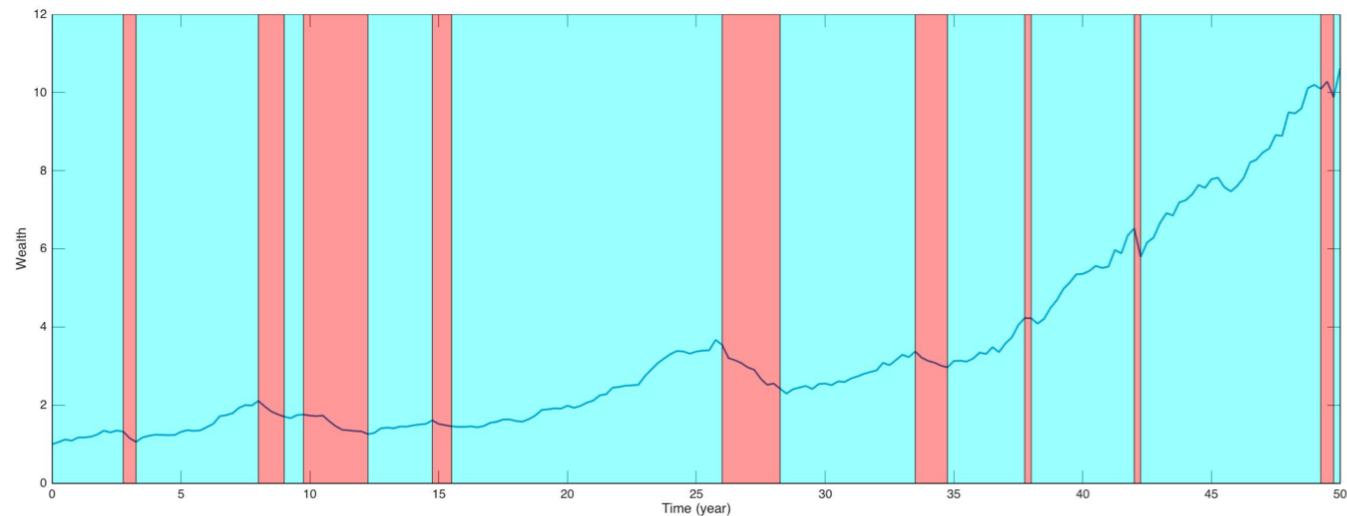
Employ these probabilities in the two-regime simulation

Multi-Period Simulation

- Single regime: select asset returns each period via a single multi-normal distribution; pay operating budget at 4% of capital (averaged over 4 years)
- Two regime: start with non-crash distribution; switch between non-crash and crash each quarter depending upon transition probability; pay operating budget the same as above
- *The two regime model more accurately projects the worst events (left tails) than the single regime approach*

Forward Looking Simulation – Sample Path

A Representative Scenario Path over the 50-Year Planning Horizon
(S&P 500 Index)



Compare Single and Two-Regime Models

Summary Statistics for Baseline Monte Carlo Simulations (4% Target Spending Target)

	Without spending-cut rule		Spending Cut by 20%	
Simulation Results	1-Regime	2-Regime	1-Regime	2-Regime
Crash Prob, 5 years	10.3%	18.4%	10.3%	18.4%
Crash Prob, 10 years	20.5%	31.8%	20.0%	31.5%
Crash Prob, 50 years	4.9%	19.9%	2.2%	13.1%
mean-5 years	1.0644	1.0780	1.0644	1.0780
mean-10 years	1.1635	1.1793	1.1692	1.1879
mean-20 years	1.3998	1.4230	1.4173	1.4514
mean-50 years	2.6630	2.6197	2.7152	2.7147
# of simulations	10000	10000	10000	10000
Average % time in "adverse"	2.88%	5.60%	2.22%	4.60%

Advantages of Reducing Spending

Panel B: Performance Statistics for 3.5% Spending Target

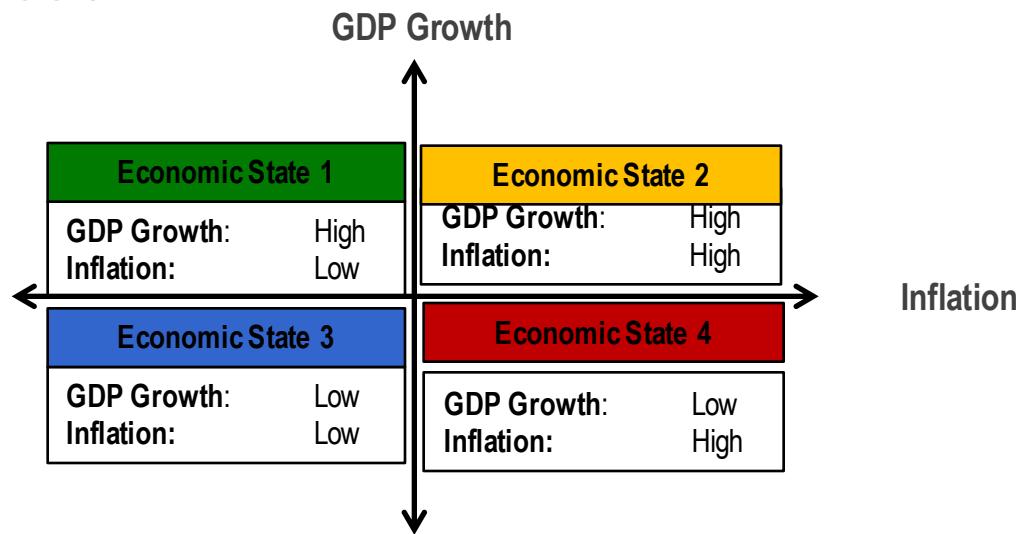
	Without spending-cut rule		Spending Cut by 25%	
Simulation Results	1-Regime	2-Regime	1-Regime	2-Regime
Crash Prob, 5 years	8.3%	16.3%	8.3%	16.3%
Crash Prob, 10 years	14.6%	26.6%	14.4%	26.4%
Crash Prob, 50 years	1.2%	8.5%	0.4%	5.6%
mean-5 years	1.0977	1.1112	1.0977	1.1112
mean-10 years	1.2421	1.2609	1.2462	1.2680
mean-20 years	1.5883	1.6267	1.6018	1.6499
mean-50 years	3.6169	3.6499	3.6585	3.7310
# of simulations	10000	10000	10000	10000
Average % time in "adverse"	1.39%	3.41%	1.06%	2.68%

Potential Advantages of Regime-Aware ALM

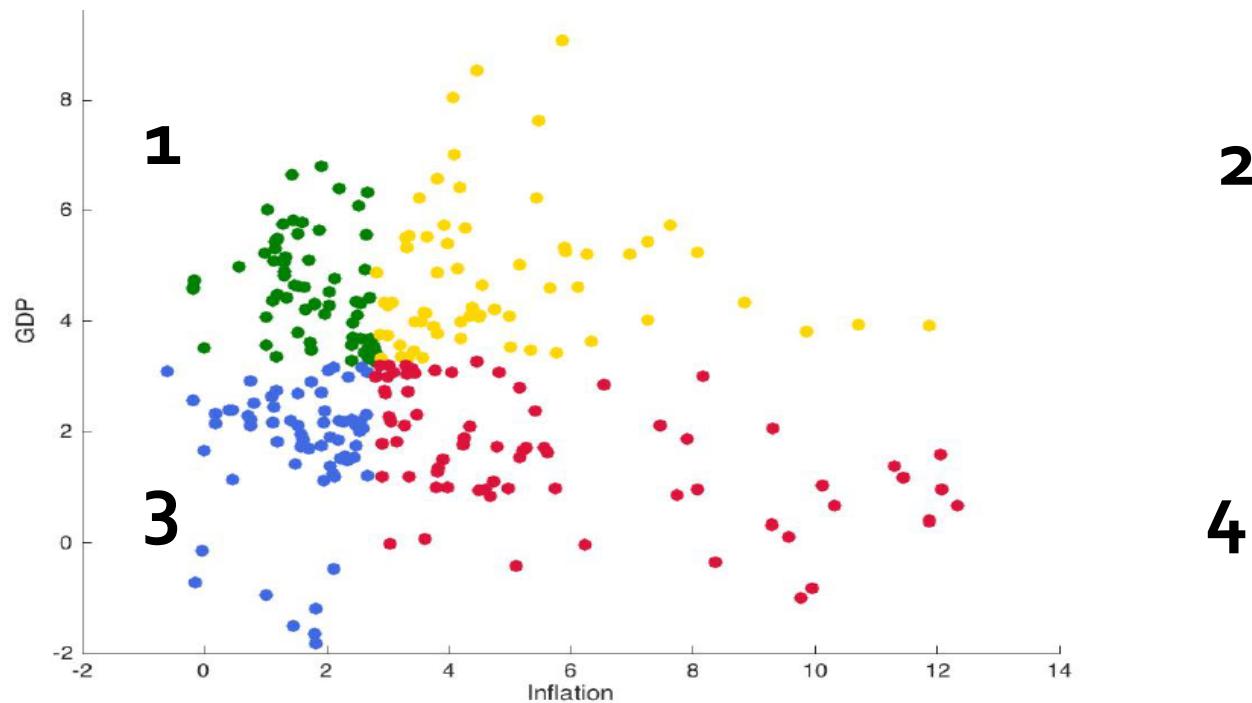
1. Consistency between asset performance and changes in liability cash flows
2. Improve estimates of downside risks
3. Enhance asset performance (possibly)
4. Streamline estimates of liability cash flows via stochastic analysis

An Illustrative Example

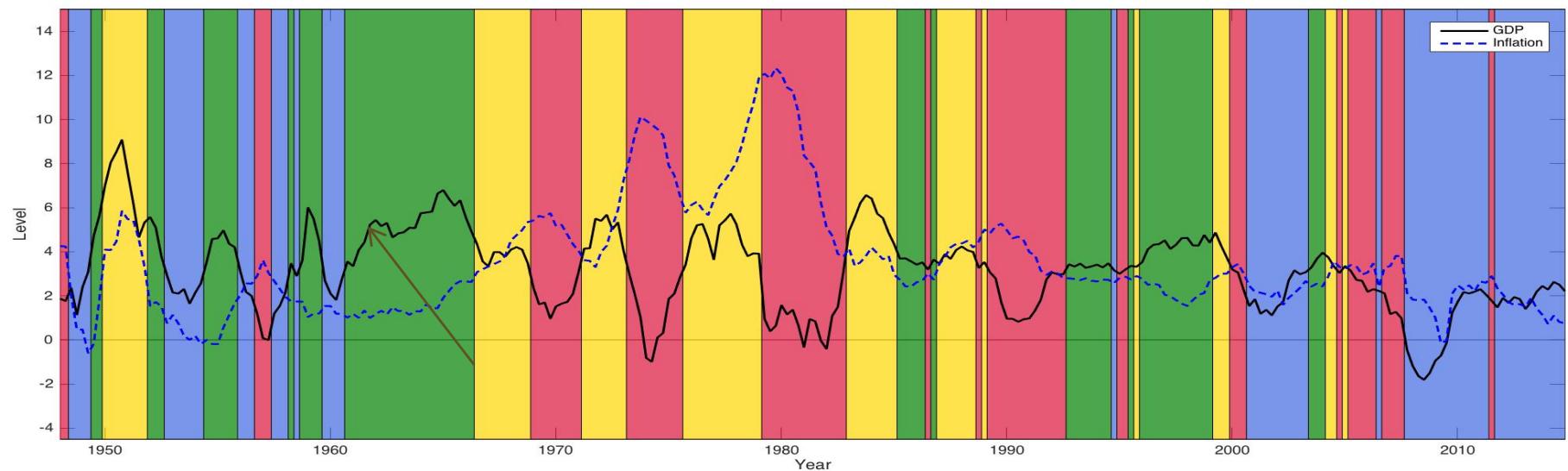
- Assume that the current state of the economy is defined by Real GDP growth and inflation:



Scatter Plot of Inflation and Real GDP 1948-2015 (4-Way Split)



Historical Patterns (time series of inflation and real GDP)



Four Regimes are Stable Across Time

Time Period 1948-2014

Frequency Quarterly

	Regime 1	Regime 2	Regime 3	Regime 4
Regime 1	0.83	0.09	0.07	0.01
Regime 2	0.05	0.83	0.00	0.13
Regime 3	0.10	0.00	0.84	0.06
Regime 4	0.04	0.07	0.08	0.80

Performance of Asset Categories

Real Returns of Major Asset Categories (1973-2015 monthly)1

Geometric Mean of Return (Annually)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
	5.8354%	4.6518%	4.0300%	3.5651%	5.3212%	2.4234%	2.9045%	0.6919%

Geometric Mean of Return (Annually)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
Regime 1	15.2834%	13.4923%	6.9086%	5.1891%	9.1384%	9.7553%	3.7975%	1.4420%
Regime 2	0.8939%	6.4105%	-0.1339%	1.4991%	3.9968%	4.7818%	1.0627%	0.7616%
Regime 3	11.0503%	10.6285%	5.1594%	7.4233%	13.7547%	-0.8822%	6.1820%	0.2001%
Regime 4	-2.8658%	-10.2222%	4.3173%	0.3051%	-4.6992%	-3.4542%	0.6718%	0.3682%

Regime 1 = growth+ and inflation-, Regime 2= growth+, inflation+, Regime 3 = growth-, inflation-, Regime 4 = growth-, inflation+

Compare Traditional MVO and Regime-Aware Approach

(see appendix for details)

Opt Portfolios	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free	Volatility
Full period	18.89%	14.70%	8.30%	26.28%	8.73%	23.10%	0.00%	0.00%	8.15%
Regime 3	0.00%	7.54%	26.28%	0.00%	27.39%	0.00%	38.79%	0.00%	8.43%

Exhibit 22
Total Return for the MVO allocation and the Regime-Aware allocation

	Full Period	Regime 3 Aware
2008	-30.25%	-10.9%
2008-2014	5.6%	43.1%

Comparing Pension Systems

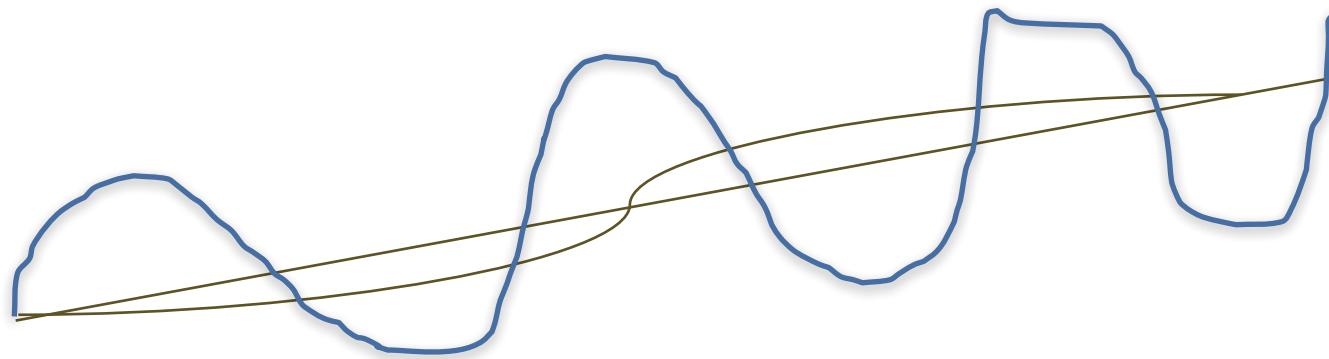
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US ⁶	25,411	131.2%
Total	41,355	67.0%⁸

Source: Willis Towers Watson and secondary sources

Difficulties Explaining Assets and Liabilities via Common Factors

1. Assets returns are driven by short and mid-term factors (interest rates, risk premium, cash flows, and micro factors such as momentum, value and so on)
2. Liabilities are driven by mid to long term factors
3. Longevity issues



Market factors (short horizon), Macro-economic factors (intermediate horizon), demographic factors (long horizon)

Macro-Economic Factors and Pension Liabilities

- Two macro-economic factors that affect liability cash flows are:
 - Economic Growth – Real Gross Domestic Product (GDP)
 - Inflation
- Impact on salaries and retirement benefits

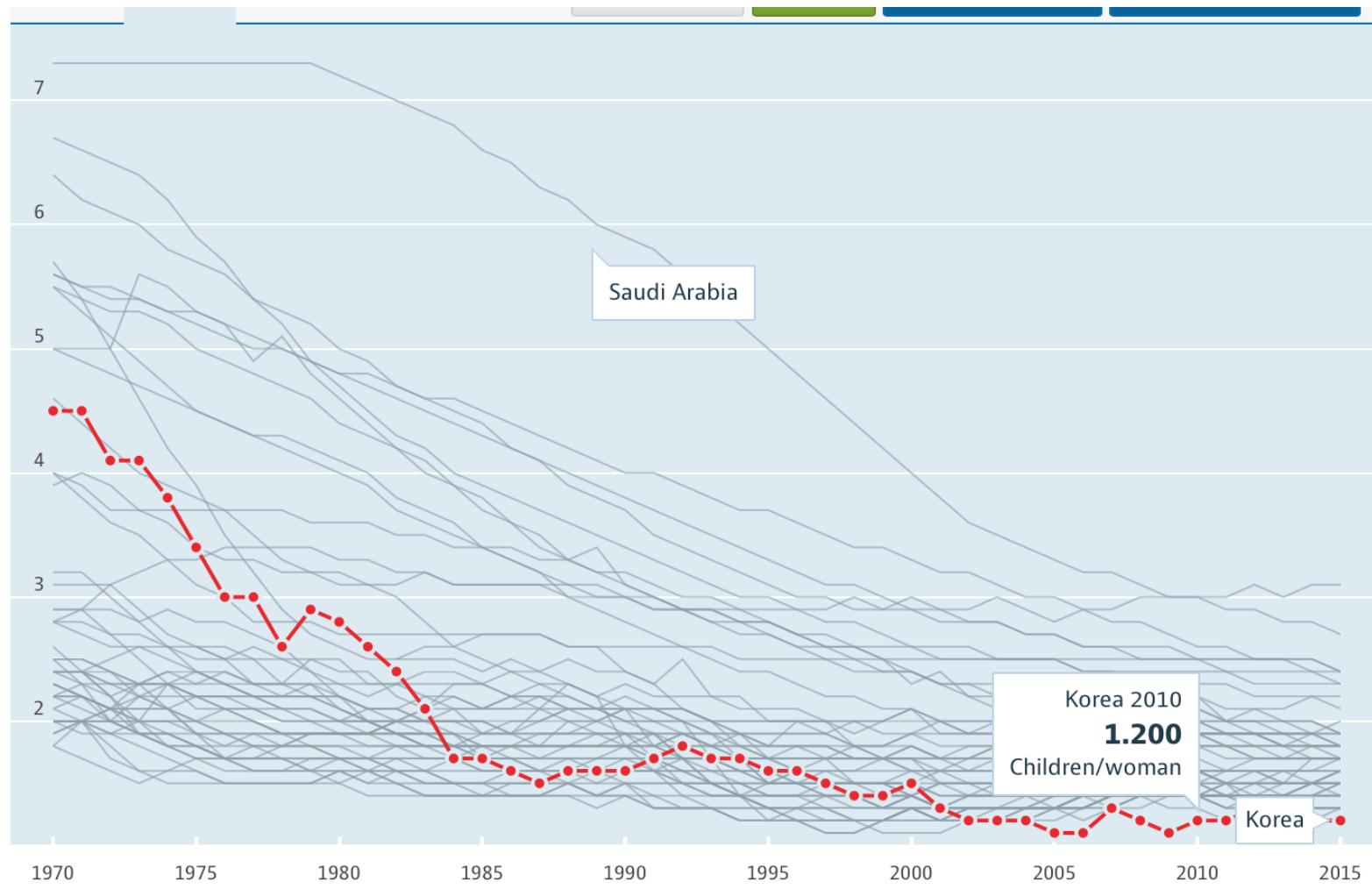
Setting the Future Goals

- It is critical to define goals that are attainable
- Compare U.S. social security versus California State Employee pension system
- For simplicity on social security
 - Assume current longevity estimates and maximum contribution (\$127k salary, \$16k per year contribution – ½ each for employee and employer)
 - Work 40 years – age 25 to 65
 - Retire 20 years – 65 to 85
 - Assume portfolio return is inflation rate and no-time value of money
 - Savings = $\$16k * 40 = \$640k$
 - Spending = $\$32k * 20 = \$640k$

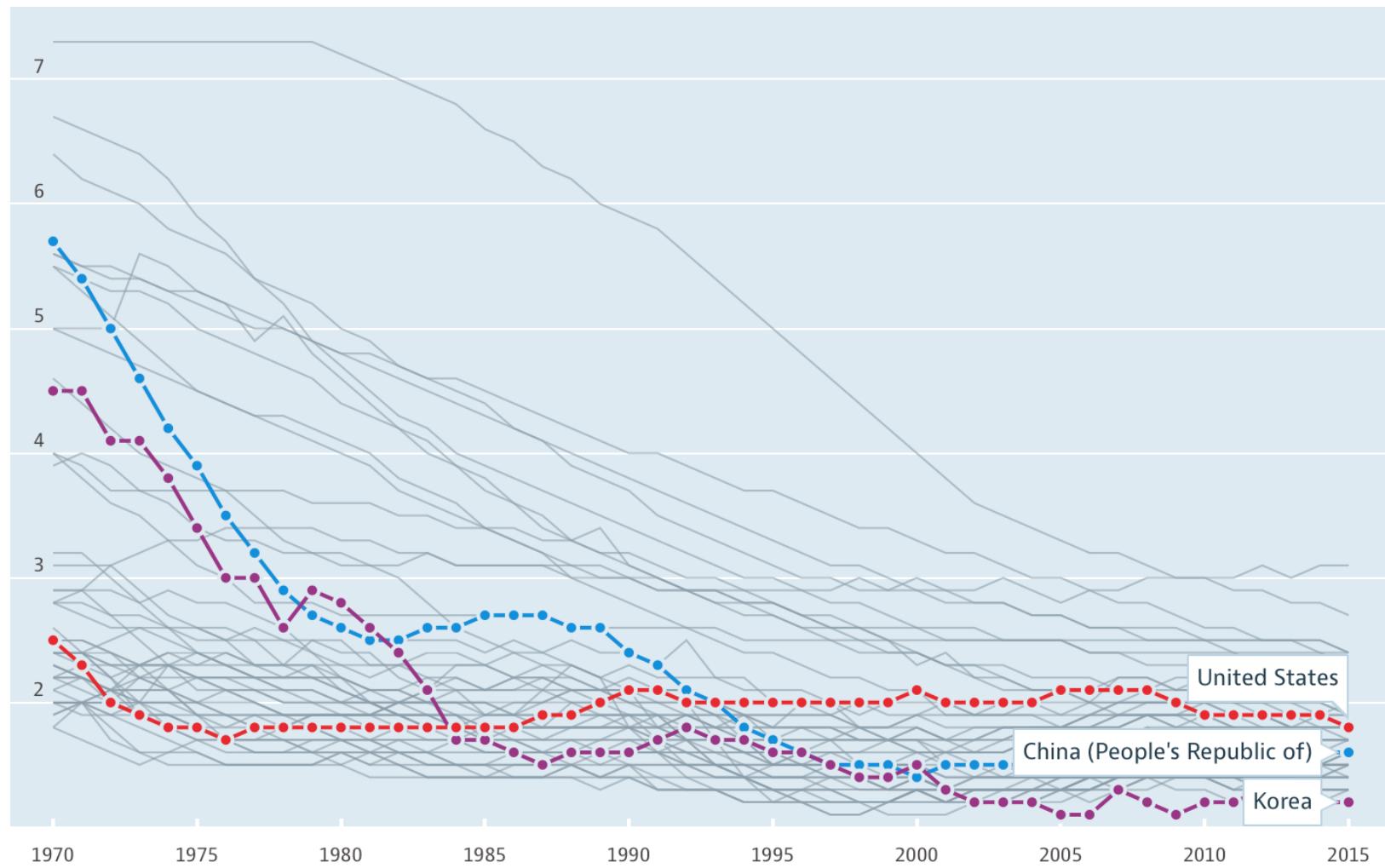
Conclusions for Social Security

- A highly conservative plan (TIPs) achieves the objectives -- 25% spending during retirement
 - If return is 3.6% real -> \$64k per year spending
 - If return is 7.2% real -> \$128k per year spending
- Of course there are longevity risks – address by delaying retirement
- And the demographic nightmare is occurring in many countries such as South Korea

Birth Rates over Time (children/woman)



Comparison of Countries – U.S., Korea, and China



State of California

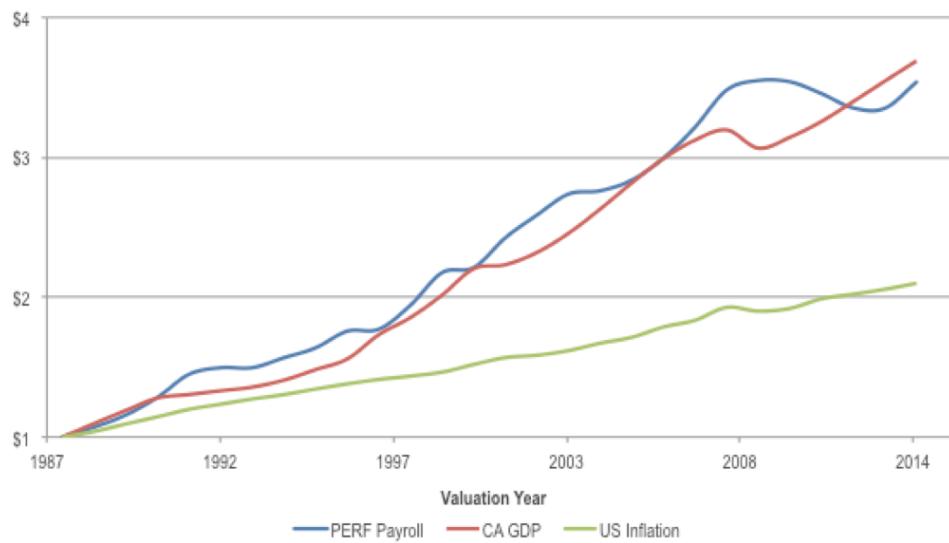
Exhibit 1 State Expenditures Since 1970

	Population	State Employees	Expenditures Per Capita	Personal Income	CPI
1970	20.0 million	181,600	\$327	\$96 B	\$1.00
2015	39.1 million	350,800	\$4400	\$2043 B	\$6.50
ratio	1.96	1.93	13.4	21.3	6.5

State of California Total Salaries

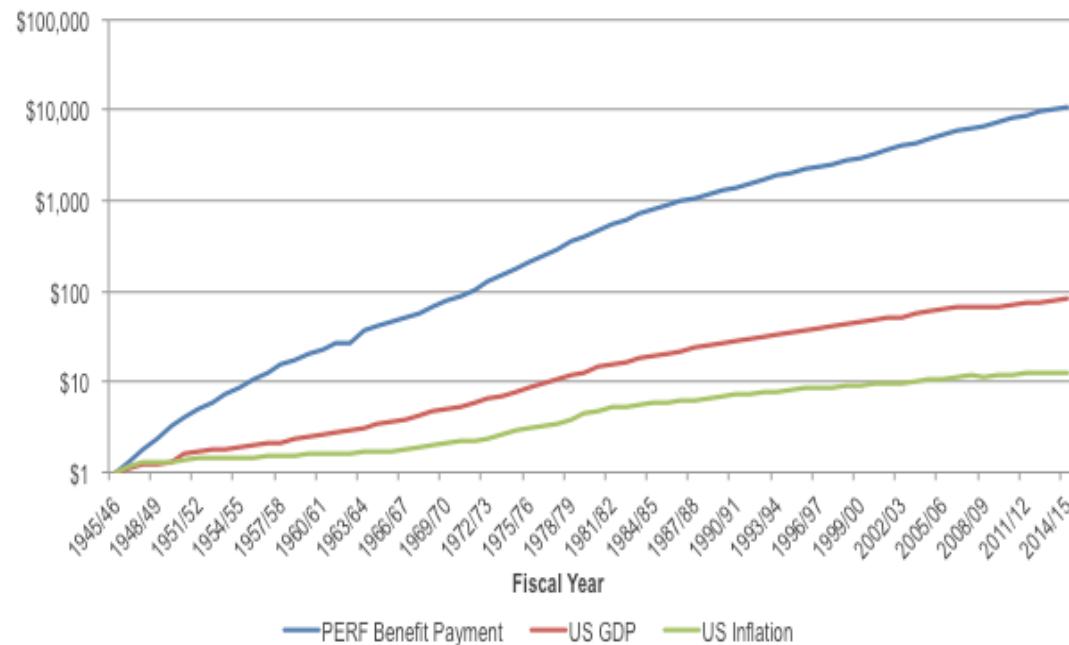
(inflation + increase in workforce ->. Underfunded plans)

Exhibit 3
PERF Payroll has increased roughly in line with California GDP (nominal)



History - Pension Payroll, GDP and US Inflation – California Employees State Pension System (CalPERS) -- \$350 Billion

- Pension Payroll and GDP showed a strong relationship



Conclusions

- U.S. pension system is largest in the world - \$25,000 Billion
 - YET: the promises are much too generous and thereby many plans are underfunded
 - Most organizations are turning to DC plans, but individuals do not have abilities to make educated investment and savings decisions
-
- **Recommendation:** increase access to social security type plans for individuals + give investors opportunities to achieve higher goals – aspirational goals.
 - Princeton/EDHEC Index



References

- Mulvey, J.M. and M. Holen, "What Can We Learn from University Endowments: Insights from Self-Reports," *Journal of Investment Consulting*, 2017.
 - "Identifying Economic Regimes: Reducing Downside Risks for University Endowments and Foundations, *Journal of Portfolio Management*, Fall 2016.
 - "A Nonparametric Smoothing Approach to Financial Market Regime Identification," Princeton University Report, J. Mulvey, H. Liu, and T. Zhang, August 2015.
-
- "Dynamic Asset Allocation for Varied Financial Markets under Regime Switching Framework," *European Journal of Operational Research*, G. Bae., W. Kim, and J. Mulvey 2013.
 - "Assisting Defined-Benefit Pension Plans," *Operations Research*, 2009 (with K. Simsek, Z. Zhang, and F. Fabozzi).
 - Willis Towers Watson, Global Pension Asset Study 2017.
 - **The Elements of Statistical Learning**, 2nd Edition, T. Hastie, R. Tibshirani, and J. Friedman (2009).

Appendix

Assets of Sovereign Wealth Funds

Country	Sovereign Wealth Fund	Assets
Norway	Government Pension Fund—Global	\$893
United Arab Emirates (Abu Dhabi)	Abu Dhabi Investment Authority	773
Saudi Arabia	SAMA Foreign Holdings	757
China	China Investment Corporation	653
China	SAFE Investment Company	568
Kuwait	Kuwait Investment Authority	548
China (Hong Kong)	Hong Kong Monetary Authority Investment Portfolio	400
Singapore	Government of Singapore Investment Corporation	320
Qatar	Qatar Investment Authority	256
Singapore	Temasek Holdings	<u>177</u>
Total		\$6,800

Trend Filtering Algorithm

A detailed formulation of the trend filtering is as follows. Let $\mathbf{Y} = (Y_1, \dots, Y_n)^T \in \mathbb{R}^n$ denote the price of a series at n evenly spaced time points. For a given integer k , the general form of a trend filtering estimator $\hat{\boldsymbol{\beta}} = (\hat{\beta}_1, \dots, \hat{\beta}_n)^T \in \mathbb{R}^n$ is defined as the solution to the following penalized optimization problem:

$$\hat{\boldsymbol{\beta}} = \underset{\boldsymbol{\beta} \in \mathbb{R}^n}{\operatorname{argmin}} \|\mathbf{Y} - \boldsymbol{\beta}\|_2^2 + \lambda \|\mathbf{D}^{(k+1)}\boldsymbol{\beta}\|_1, \quad (2.1)$$

where $\lambda \geq 0$ is a regularization parameter, and $\mathbf{D}^{(k+1)} \in \mathbb{R}^{(n-k-1) \times n}$ denotes the operator for computing the $(k+1)$ -th order discrete derivative. For example, when $k=0$ and $k=1$,

$$\mathbf{D}^{(1)} = \begin{pmatrix} 1 & -1 & 0 & \dots & 0 & 0 \\ 0 & 1 & -1 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 1 & -1 \end{pmatrix}, \quad \mathbf{D}^{(2)} = \begin{pmatrix} 1 & -2 & 1 & \dots & 0 & 0 \\ 0 & 1 & -2 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & -2 & 1 \end{pmatrix}$$

so that $\|\mathbf{D}^{(1)}\boldsymbol{\beta}\|_1 = \sum_{i=1}^{n-1} |\beta_i - \beta_{i+1}|$, and $\|\mathbf{D}^{(2)}\boldsymbol{\beta}\|_1 = \sum_{i=1}^{n-2} |\beta_i - 2\beta_{i+1} + \beta_{i+2}|$.

Trend Filtering Algorithm

With different choices of k , the solution takes on different structures. When $k = 0$, the solution to (2.1) is a piecewise step function. When $k = 1$, the solution is piecewise linear. When $k = 2$, the solution is piecewise quadratic, and so on. To see the intuition of piecewise linear when $k = 1$: Eq. (2.1) is in a form of generalized lasso (least absolute shrinkage and selection operator) problem (Tibshirani et al., 2011, 2012):

$$\hat{\beta} = \underset{\beta \in \mathbb{R}^p}{\operatorname{argmin}} \|Y - X\beta\|_2^2 + \lambda \|H\beta\|_1, \quad (2.2)$$

where $X \in \mathbb{R}^{n \times p}$, and $H \in \mathbb{R}^{m \times p}$. When $m = p$ and H is the identity matrix, then (2.2) becomes the regular lasso estimator (Tibshirani, 1996). The lasso was originally proposed

Trend Filtering Algorithm

to handle high dimensional sparse regression and variable selection problems. Its solution has the remarkable property of being sparse (i.e., many entries of the solution vector are zero), while the optimization problem remains convex and efficient to solve. There is a straightforward geometric interpretation for the sparse property of lasso. By Langrange multiplier theory, the formulation of lasso is equivalent to a constraint optimization problem:

$$\hat{\beta} = \underset{\beta \in \mathbb{R}^d}{\operatorname{argmin}} \frac{1}{n} \sum_{i=1}^n (Y_i - X_i^T \beta)^2 \quad \text{s.t.} \quad \|\beta\|_1 \leq \mu,$$

for some $\mu \in \mathbb{R}$ and X_i is the i -th row of X . Similar transformation is true for ℓ_0 - and ridge regression, which uses $\|\cdot\|_0$ and $\|\cdot\|_2$ norm as regularization function.

Performance of Asset Categories

Real Returns of Major Asset Categories (1973-2015 monthly)1

Geometric Mean of Return (Annually)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
	5.8354%	4.6518%	4.0300%	3.5651%	5.3212%	2.4234%	2.9045%	0.6919%

Geometric Mean of Return (Annually)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
Regime 1	15.2834%	13.4923%	6.9086%	5.1891%	9.1384%	9.7553%	3.7975%	1.4420%
Regime 2	0.8939%	6.4105%	-0.1339%	1.4991%	3.9968%	4.7818%	1.0627%	0.7616%
Regime 3	11.0503%	10.6285%	5.1594%	7.4233%	13.7547%	-0.8822%	6.1820%	0.2001%
Regime 4	-2.8658%	-10.2222%	4.3173%	0.3051%	-4.6992%	-3.4542%	0.6718%	0.3682%

Regime 1 = growth+ and inflation-, Regime 2= growth+, inflation+, Regime 3 = growth-, inflation-, Regime 4 = growth-, inflation+

Equity Micro-Factor Performance

Real Return of Equity Micro-Factors over Four Regimes – 1970-2015 monthly

	Geometric Mean of Return (Quarterly)				High	Low	High	Low	High	Low
	High Value	Low Value	High Vol	Low Vol	Investment	Investment	Profitability	Profitability	Momentum	Momentum
Regime 1	5.36%	4.20%	4.73%	4.08%	4.37%	4.99%	5.42%	4.30%	6.36%	3.53%
Regime 2	3.40%	0.97%	1.91%	2.19%	1.44%	2.49%	1.96%	1.82%	2.51%	1.23%
Regime 3	2.86%	0.97%	1.01%	2.19%	0.78%	2.54%	2.39%	1.23%	1.12%	1.88%
Regime 4	2.78%	0.86%	2.14%	1.63%	1.63%	2.26%	1.82%	2.01%	2.48%	1.22%

Modest advantage for long-only micro factors

→ long/short better diversification

Comparing Traditional MVO and Regime-Aware Approach

Inputs to a Markowitz Portfolio Model January 1973-December 2007, Real Monthly Returns

Geometric Mean of Return (Monthly)

U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
0.4803%	0.4870%	0.2995%	0.2806%	0.4222%	0.4727%	0.2507%	0.0968%

Conditional Value at Risk (Monthly)

U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
-10.0627%	-10.6460%	-5.9763%	-4.2674%	-10.7883%	-11.5732%	-6.8351%	-0.3874%

Volatility (Monthly)

U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
0.044211	0.047802	0.030183	0.021728	0.045102	0.055778	0.028858	0.002063

Sharpe Ratio (Monthly)

U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
0.086752	0.081635	0.067182	0.084619	0.072162	0.067397	0.053333	0.000000

Correlation

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
U.S.Equity	1.000000	0.574599	0.211152	0.236128	0.565631	0.005373	0.114511	0.011030
Intl.Equity	0.574599	1.000000	0.134243	0.204139	0.387313	0.062845	0.070981	0.022911
U.S.Treasury	0.211152	0.134243	1.000000	0.741640	0.238839	-0.049413	0.481781	0.174982
Corp.Bond	0.236128	0.204139	0.741640	1.000000	0.288717	-0.050624	0.387406	0.197569
Real Estate	0.565631	0.387313	0.238839	0.288717	1.000000	-0.041503	0.214121	-0.034849
Commodity	0.005373	0.062845	-0.049413	-0.050624	-0.041503	1.000000	0.101095	-0.015984
TIPS	0.114511	0.070981	0.481781	0.387406	0.214121	0.101095	1.000000	0.096095
Risk Free	0.011030	0.022911	0.174982	0.197569	-0.034849	-0.015984	0.096095	1.000000

Comparing Traditional MVO and Regime-Aware Approach

Performance of Assets Under the Four Regimes January 1973 to December 2007

Geometric Mean of Return (Monthly)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
Regime 1	1.1820%	1.1216%	0.4543%	0.3829%	0.6396%	1.1698%	0.3033%	0.1574%
Regime 2	0.0742%	0.5191%	-0.0112%	0.1036%	0.3271%	0.3900%	0.0881%	0.0632%
Regime 3	0.5105%	0.6285%	0.6368%	0.5022%	0.7955%	0.0284%	0.6426%	0.0943%
Regime 4	0.2245%	-0.2829%	0.2715%	0.2300%	0.0574%	0.1740%	0.1165%	0.0759%

Geometric Mean of Return (Annually)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
Regime 1	15.1435%	14.3208%	5.5896%	4.6924%	7.9512%	14.9772%	3.7006%	1.9056%
Regime 2	0.8939%	6.4105%	-0.1339%	1.2503%	3.9968%	4.7818%	1.0627%	0.7616%
Regime 3	6.3007%	7.8078%	7.9149%	6.1962%	9.9747%	0.3409%	7.9896%	1.1377%
Regime 4	2.7279%	-3.3423%	3.3074%	2.7946%	0.6906%	2.1084%	1.4074%	0.9152%

Conditional Value at Risk (Monthly)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
Regime 1	-5.4773%	-7.0090%	-4.7597%	-2.6150%	-6.5103%	-8.4059%	-3.8300%	-0.1358%
Regime 2	-11.7270%	-11.3778%	-4.9681%	-3.8713%	-10.8949%	-11.0935%	-7.0140%	-0.2242%
Regime 3	-9.6752%	-11.2141%	-6.6980%	-2.4451%	-8.7136%	-11.6188%	-3.2554%	-0.1359%
Regime 4	-10.8876%	-11.9761%	-7.0830%	-6.8815%	-14.0816%	-14.5317%	-8.9285%	-0.5232%

Conditional Value at Risk (Annually)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
Regime 1	-49.1331%	-58.1890%	-44.3009%	-27.2381%	-55.4172%	-65.1333%	-37.4140%	-1.6173%
Regime 2	-77.6161%	-76.5303%	-45.7459%	-37.7359%	-74.9485%	-75.6105%	-58.2161%	-2.6576%
Regime 3	-70.5094%	-76.0045%	-56.4795%	-25.6995%	-66.5133%	-77.2848%	-32.7766%	-1.6182%
Regime 4	-74.9241%	-78.3626%	-58.5865%	-57.4960%	-83.8179%	-84.8063%	-67.4469%	-6.1009%

Volatility (Monthly)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
Regime 1	0.036762	0.040043	0.024361	0.014356	0.034961	0.048538	0.018276	0.001563
Regime 2	0.044392	0.045772	0.026912	0.018935	0.042762	0.057525	0.027643	0.001952
Regime 3	0.042659	0.050499	0.032279	0.014211	0.036556	0.052504	0.021081	0.001437
Regime 4	0.051203	0.054603	0.036895	0.032596	0.060046	0.062507	0.041122	0.002771

Sharpe Ratio (Monthly)

	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free
Regime 1	0.278706	0.240775	0.121850	0.157036	0.137920	0.208577	0.079799	0.000000
Regime 2	0.002465	0.099597	-0.027649	0.021309	0.061706	0.056803	0.009002	0.000000
Regime 3	0.097557	0.105774	0.168056	0.287058	0.191804	-0.012562	0.260078	0.000000
Regime 4	0.029019	-0.065716	0.053010	0.047248	-0.003093	0.015691	0.009871	0.000000

Compare Traditional MVO and Regime-Aware Approach

Opt Portfolios	U.S.Equity	Intl.Equity	U.S.Treasury	Corp.Bond	Real Estate	Commodity	TIPS	Risk Free	Volatility
Full period	18.89%	14.70%	8.30%	26.28%	8.73%	23.10%	0.00%	0.00%	8.15%
Regime 3	0.00%	7.54%	26.28%	0.00%	27.39%	0.00%	38.79%	0.00%	8.43%

Exhibit 22
Total Return for the MVO allocation and the Regime-Aware allocation

	Full Period	Regime 3 Aware
2008	-30.25%	-10.9%
2008-2014	5.6%	43.1%