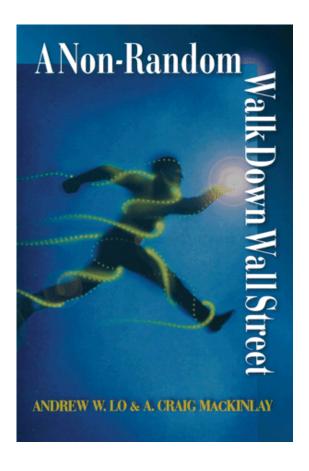




#### 1. DELIVERY OF MANUSCRIPT

You agree to deliver the complete manuscript ready for copyediting, including a preface, other front matter, and back matter, together with camera-ready copy for all illustrations, maps, charts, drawings, or other material (except index) to be included in the Work, not later than April 15, 2008) or a later date designated in writing by the Press (the "Delivery Date"), time of delivery being of the essence. The final manuscript submitted to the Press will consist of no more than 80,000 words in length (including text, notes, and bibliography), 20 illustrations, and 10 color plates, and we reserve the right to reject the manuscript if the manuscript exceeds this limit.





In a fashion analogous to Theorem 6.3.1, the behavior of  $Q_n$  under long-range dependent alternatives may now be derived in several steps using Lemmas A.2, A.3, and Theorem 6.3.2:

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**Theorem 6.3.3.** Let  $\{\epsilon_i\}$  be a zero-mean stationary Gaussian stochastic process with autocovariance function y such that

$$\gamma_k \sim \begin{cases} k^{2H-2}L(k) & \text{for } H \in (\frac{1}{2}, 1) \text{ or,} \\ -k^{2H-2}L(k) & \text{for } H \in (0, \frac{1}{2}) \end{cases}$$
 as  $k \to \infty$  (6.3.14)

where L(k) is a slowly varying function at infinity. Then as n and q increase without bound such that  $(q/n) \rightarrow 0$ , we have:

(a) 
$$\underset{1 \le k \le n}{\operatorname{Max}} \ \frac{1}{\sigma_n} \sum_{j=1}^k (X_j - \overline{X}_n) \quad \Rightarrow \quad \underset{0 \le \tau \le 1}{\operatorname{Max}} \ W_H^{\circ}(\tau) \equiv M_H^{\circ},$$

(b) 
$$\min_{1 \le k \le n} \frac{1}{\sigma_n} \sum_{i=1}^k (X_i - \overline{X}_n) \implies \min_{0 \le \tau \le 1} W_H^{\circ}(\tau) = \tau$$

(c) 
$$R_n \equiv \frac{\hat{\sigma}_n(q)\sqrt{n}}{N} \cdot \frac{1}{N} Q_n \Rightarrow M_H^\circ - m_H^\circ \equiv V_H$$

(d) 
$$a_n \equiv \frac{\sigma_n}{\hat{\sigma}_n(q)\sqrt{n}} \xrightarrow{p} \begin{cases} \infty & for H \in (\frac{1}{2}, 1) \\ 0 & for H \in (0, \frac{1}{2}) \end{cases}$$

$$\begin{array}{llll} \text{(b)} & & \displaystyle \min_{1 \le k \le n} \frac{1}{\sigma_n} \sum_{j=1}^k (X_j - \overline{X}_n) & \Rightarrow & \displaystyle \min_{0 \le \tau \le 1} W_H^\sigma(\tau) = m_H^\sigma, \\ \text{(c)} & & R_n = \frac{\hat{\sigma}_n(q) \sqrt{n}}{\sigma_n} \cdot \frac{1}{\sqrt{n}} Q_n & \Rightarrow & M_H^\sigma - m_H^\sigma \equiv V_H, \\ \text{(d)} & & a_n \equiv \frac{\sigma}{\hat{\sigma}_n(q) \sqrt{n}} & \stackrel{p}{\to} & \begin{cases} \text{for } H \in (\frac{1}{2}, 1), \\ 0 & \text{for } H \in (0, \frac{1}{2}), \end{cases} \\ \text{(e)} & & \frac{1}{\sqrt{n}} Q_n = a_n R_n & \stackrel{p}{\to} & \begin{cases} \infty & \text{for } H \in (\frac{1}{2}, 1), \\ 0 & \text{for } H \in (0, \frac{1}{2}), \end{cases} \end{array}$$

where  $\hat{\sigma}_n(q)$  is defined in (6.3.6),  $\sigma_n$  is defined in Theorem 6.3.2, and  $W_H^\circ(\tau)\equiv W_H(\tau)-\tau\,W_H(1)$ .

Theorem 6.3.3 shows that the modified rescaled range test is consistent against a class of long-range dependent stationary Gaussian alternatives. In the presence of positive strong dependence, the R/S statistic diverges in probability to infinity; in the presence of negative strong dependence, it converges in probability to zero. In either case, the probability of rejecting the null hypothesis approaches unity for all stationary Gaussian stochastic processes satisfying (6.3.14), a broad set of alternatives that includes all fractionally-differenced Gaussian ARIMA(p, d, q) models with  $d \in (-\frac{1}{2}, \frac{1}{2})$ .

From (a) and (b) of Theorem 6.3.3 it is apparent that the normalized population rescaled,  $R_n/\sqrt{n}$ , converges to zero in probability. Therefore,

 $^{20}$ Although it is tempting to call  $W_H^o(\tau)$  a "fractional Brownian bridge," this is not the most natural definition despite the fact that it is "tied down." See Jonas (1983, Chapter 3.3) for a



# Markets are efficient



People behave irrationally

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Slide 4





Efficient Markets Rational Expectations

Artificial Intelligence Bounded Rationality

Evolutionary Biology Ecology



Behavioral Finance Psychology

Cognitive Neurosciences

Adaptive Markets Hypothesis





## Summary



- Traditional investment framework is flawed
- Not wrong, but incomplete (physics envy)
- Stable environment  $\Rightarrow$  stable investment policies (EMH)
- Dynamic environment ⇒ dynamic investment policies (AMH)
- The current environment is highly dynamic
- We must adapt to changing market conditions
- "it's the economy, stupid"
- The Adaptive Markets Hypothesis provides a framework for investing, risk management, financial regulation, and more



In the beginning...

$$R_{it} = \alpha_i + \beta_i F_t + \epsilon_{it}$$

#### Implications:

- Correlation matters; diversification
- Benchmarks, performance attribution
- Passive investing
- Indexation, hedging, portable alpha
- Risk budgeting
- Framework for fiduciary duties





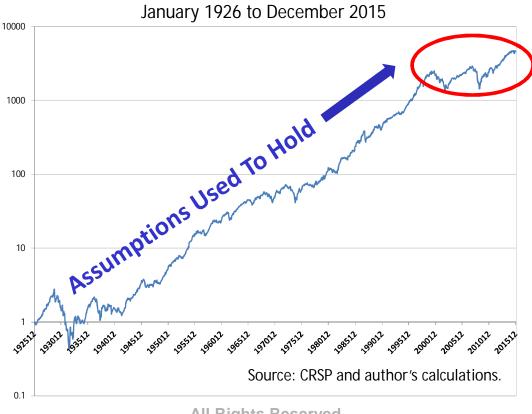
#### But This Framework Requires Several Key Assumptions:

- Relationship is linear
- Relationship is static across time and circumstances
- Parameters can be accurately estimated
- Investors behave rationally
- Markets are stationary (static probability laws)
- Markets are efficient

What If Some of These Assumptions Don't Hold?



Cumulative Return of S&P 500 (log scale)



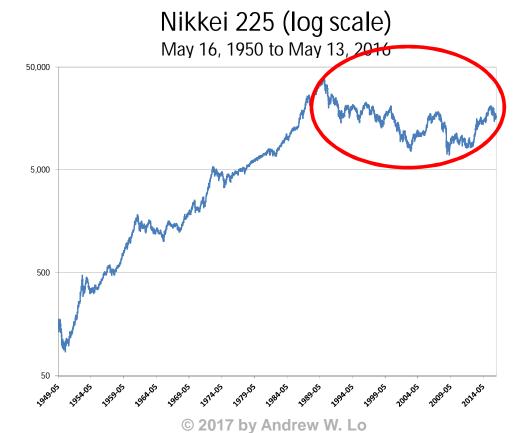
But Do They

Still Hold

Today??

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#### Have Alternatives Become Irrelevant?



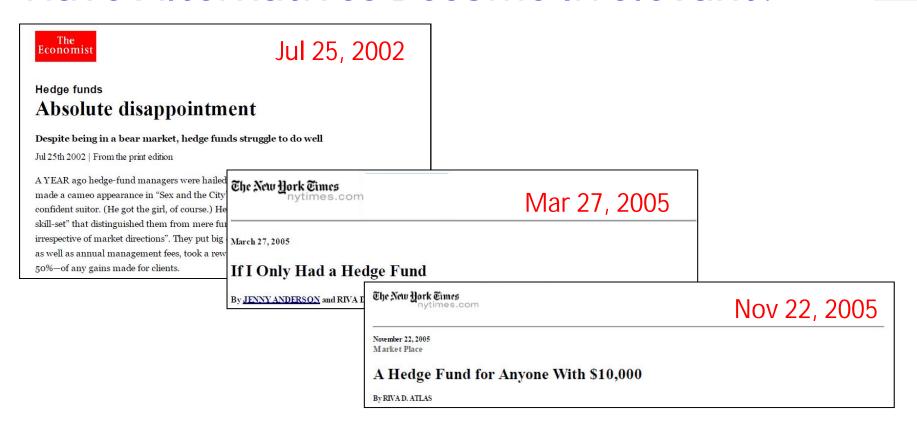






#### Have Alternatives Become Irrelevant?

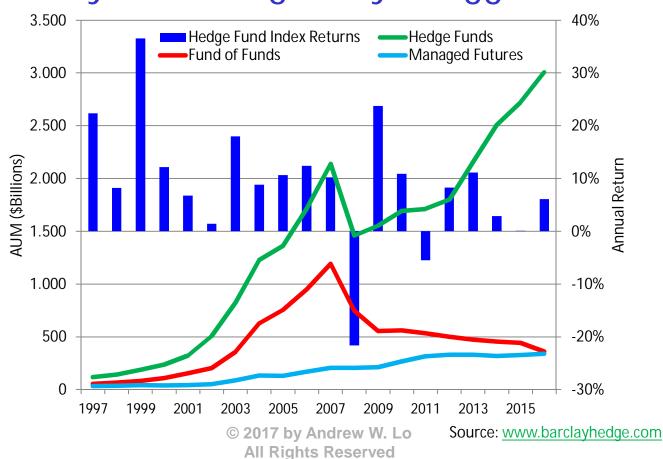




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#### MIT LFE

## "News of my death is greatly exaggerated..."



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## Pre- and Post-Crisis Hedge Fund Performance





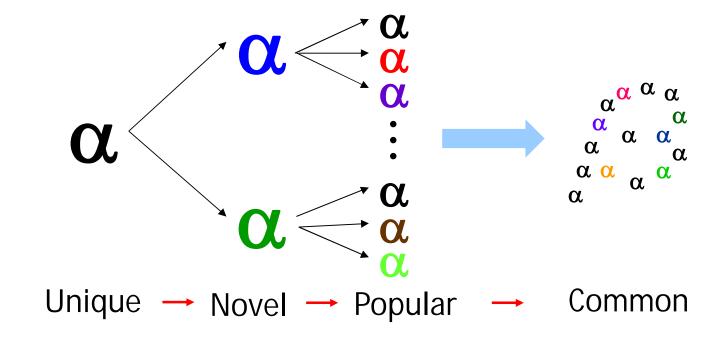
Category	# Fund- Months	Ann. Mean (%)	Ann. SD (%)	Sharpe Ratio	Sortino Ratio	Skew.	Kurt.	MaxDD (%)	Corr. to S&P 500 (%)	<b>ρ</b> <sub>1</sub> (%)	Box-Q(3) p-value (%)
January 1996 to December 2006											
Convertible Arbitrage	7,827	8.1	4.3	0.95	1.53	-1.25	8.63	-8.70	42.9	45.9	0.0
Dedicated Short Bias	1,384	-2.3	18.8	-0.31	-0.58	0.59	4.17	-42.29	-76.8	8.9	19.1
Emerging Markets	12,673	11.6	15.7	0.47	0.69	-1.61	10.51	-49.26	58.5	28.0	0.8
Equity Market Neutral	11,537	6.5	3.0	0.82	1.82	2.05	16.06	-2.21	3.0	-11.7	22.4
Event Driven	18,565	9.4	5.2	1.02	1.55	-2.02	13.89	-12.56	54.7	32.3	0.2
Fixed Income Arbitrage	7,749	6.8	3.7	0.75	0.95	-3.56	24.53	-13.69	-1.0	42.7	0.0
Global Macro	8,948	4.7	6.1	0.14	0.26	0.46	4.17	-14.24	21.7	2.5	41.6
Long/Short Equity Hedge	69,160	11.1	9.8	0.71	1.33	0.15	5.31	-18.52	68.8	18.7	16.7
Managed Futures	13,761	5.0	9.8	0.11	0.20	0.14	2.97	-16.34	-8.7	0.0	72.9
Multi-Strategy	8,100	8.5	5.2	0.85	1.43	-0.73	5.16	-6.67	49.1	0.1	65.1
Fund of Funds	55,507	6.6	6.4	0.41	0.68	-U 33	6.51	12 07	F2 2	າາ າ	5.0
All Single Manager Funds	163,702	8.7	6.5	0.72	1.28	-0.26	5.46	-10.95	65.2	19.2	13.1
			January 2010 to December 2014								
Convertible Arbitrage	3,940	3.0	5.7	0.52	0.96	-0.07	2.67	-10.20	50.4	10.6	63.6
Dedicated Short Bias	571	-1.5	7.3	-0.21	-0.33	-0.40	2.94	-22.56	-59.1	10.2	66.5
Emerging Markets	22,401	0.4	8.5	0.04	0.06	-0.67	3.84	-16.10	70.6	7.9	41.9
Equity Market Neutral	8,930	3.9	2.4	1.59	2.97	-0.69	4.00	-3.35	81.6	22.4	25.1
Event Driven	11,465	5.0	4.9	1.01	1.78	-0.70	3.04	-7.66	77.1	20.1	20.2
Fixed Income Arbitrage	7,202	5.0	1.7	2.90	6.29	-0.94	4.23	-1.03	54.1	-10.0	68.3
Global Macro	16,824	3.7	2.5	1.46	3.29	0.12	3.54	-2.03	63.2	9.7	68.7
Long/Short Equity Hedge	66,758	4.7	6.2	0.73	1.27	-0.54	3.42	-10.67	89.0	11.1	64.2
Managed Futures	23,471	2.8	7.5	0.36	0.71	0.11	2.28	-14.48	25.4	-12.4	78.4
Multi-Strategy	57,505	5.2	2.5	2.06	4.06	-0.87	5.52	-3.20	80.9	16.3	23.2
Fund of Funds	139 161	17	3.5	0.46	0.78	-0.55	2 73	-7 42	79.3	11.6	59 1

Source: Getmansky, Lee, Lo (2015, Table 14)

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## Hedge-Fund Strategy Life Cycle





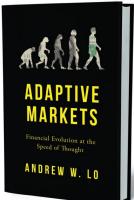




"Nothing makes sense in biology except in the light of evolution," Dobzhansky (1973)

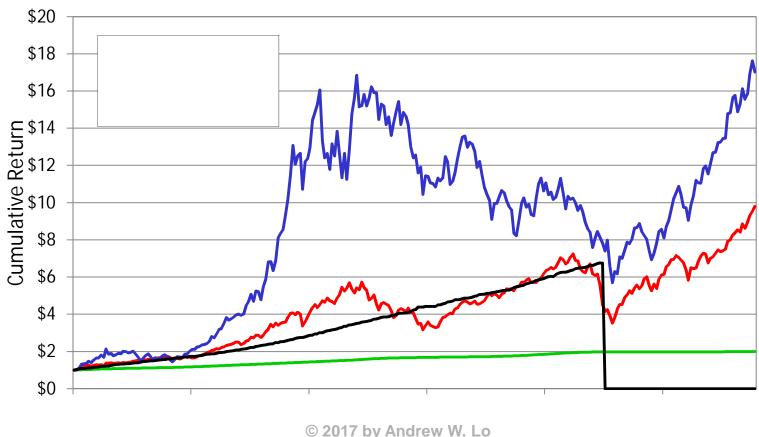
"Nothing makes sense in the hedge fund industry except in the light of the Adaptive Markets Hypothesis," Lo (2017)

- 1. Individuals act in their own self-interest
- 2. Individuals make mistakes ("satisfice")
- 3. Individuals learn and adapt (heuristics)
- 4. Competition drives adaptation and innovation
- 5. Evolution determines market dynamics



## What Do Investors Want?





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Slide 20

## Risk Perception and Adaptive Behavior



The	Eff	ects	of	Auto	mobile
Safe	ty	Regu	ıla	tion	

Sam Peltzman

University of Chicago

Technological studies imply that annual highway deaths would be 20 percent greater without legally mandated installation of various safety devices on automobiles. However, this literature ignores offsetting effects of nonregulatory demand for safety and driver response to the devices. This article indicates that these offsets are virtually complete, so that regulation has not decreased highway deaths. Time-series (but not cross-section) data imply some saving of auto occupants' lives at the expense of more pedestrian deaths and more nonfatal accidents, a pattern consistent with optimal driver response to regulation.

Journal of Political Economy 83(1975), 677–726.





## Automobile Safety Regulation and the Incentive to Drive Recklessly: Evidence from NASCAR

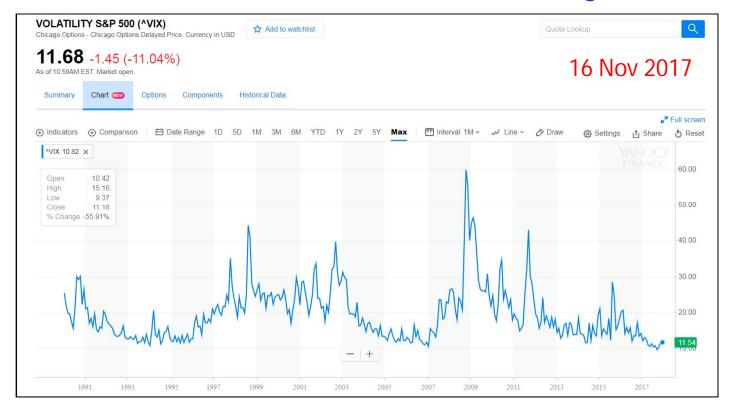
Russell S. Sobel\* and Todd M. Nesbit†

When safety regulation makes automobiles safer, drivers may drive more recklessly, partially or completely offsetting effects on the overall level of safety. Evidence of these offsetting effects has been hard to find, however, primarily because of the aggregate nature of accident data. In this paper we explore how changes in the safety of automobiles used in the National Association for Stock Car Auto Racing (NASCAR) has altered the incentive of drivers to drive recklessly. This unique data set allows more accurate and objective measurement of the necessary variables to test for these effects at a microlevel. Our results strongly support the presence of these offsetting behavioral effects.

Southern Economic Journal 74(2007), 71–84.

## Implications for the Current Ecosystem









#### **Efficient Markets**

- Long-only constraint
- Diversify across stocks and bonds
- Market-cap-weighted indexes
- Manage risk via asset allocation
- Alpha vs. market beta
- Markets are efficient
- Equities in the long run

#### **Adaptive Markets**

- Long/short strategies
- Diversify across more asset classes and strategies
- Passive transparent indexes
- Manage risk via active volatility scaling algorithms
- Alphas ⇒ multiple betas
- Markets are adaptive
- "In the long run we're all dead," but make sure the short run doesn't kill you first





#### **Efficient Markets**

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Market-cap-weighted portfolio?

Jack Bogle (1997) on the Origins of the Vanguard Index Trust:

The basic ideas go back a few years earlier. In 1969–1971, Wells Fargo Bank had worked from academic models to develop the principles and techniques leading to index investing. John A. McQuown and William L. Fouse pioneered the effort, which led to the construction of a \$6 million index account for the pension fund of Samsonite Corporation. With a strategy based on an equal-weighted index of New York Stock Exchange equities, its execution was described as "a nightmare". The strategy was abandoned in 1976, replaced with a market-weighted strategy using the Standard & Poor's 500 Composite Stock Price Index. The first such models were accounts run by Wells Fargo for its own pension fund and for Illinois Bell.





- Market-cap weighting requires little trading
- "Buy-and-hold" portfolio
- What if trading were cheaper, faster, and automatable?

In.dex \'in- deks\ noun

An index is any portfolio strategy satisfying three properties: (1) it is completely transparent; (2) it is investable; and (3) it is totally systematic.



Value-weighted average?

Equal-weighted average?

Target-date fund?

FHFA House Price Index?

Hedge Fund Index?

Trend-following futures?

Risk-managed large-cap core?

Yes	No	Maybe







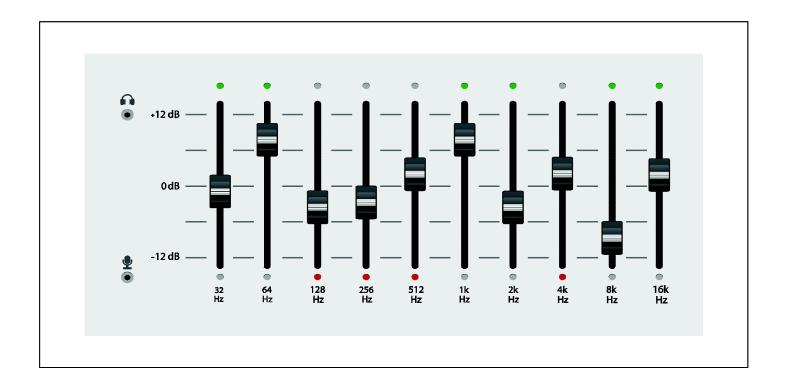




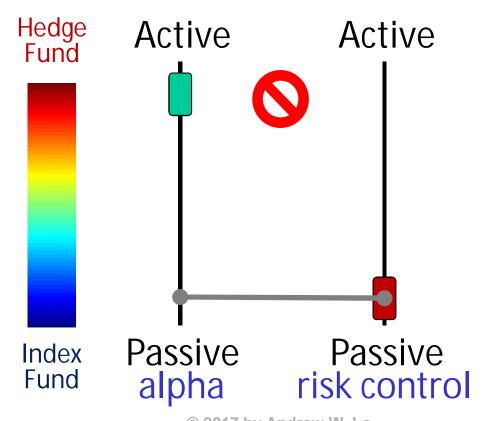












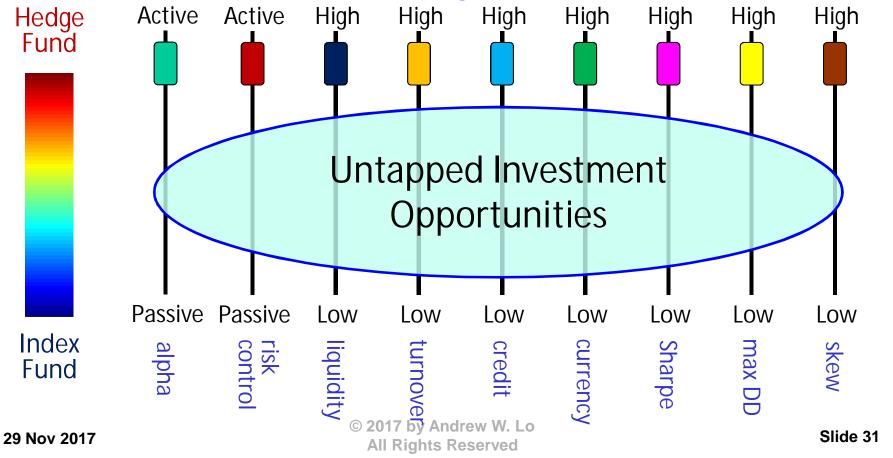
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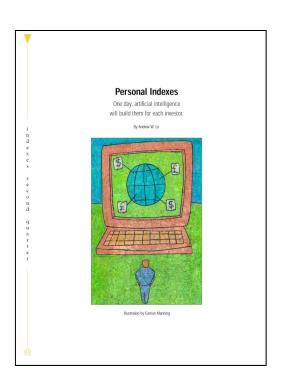


## The Opportunity: Precision Indexes

- Instead of the DowJones30®, FTSE100®, or S&P500®, imagine investing in the:
  - RichardZeckhauser30<sup>®</sup>, ArnieWood100<sup>®</sup>, or LarrySummers500<sup>®</sup>
- Imagine if such portfolios took into account income, expenses, age, health, taxes, and behavior
- Imagine if such portfolios were automated
- We have the hardware and software; we need the algorithms

#### This Idea Is Not New





"Artificial intelligence and active management are not at odds with indexation, but instead imply a more sophisticated set of indexes and portfolio management policies for the typical investor, something each of us can look forward to, perhaps within the next decade."

- Andrew W. Lo, Journal of Indexes Q2, 2001

## So What's Missing?



## ...Not Artificial Intelligence

## **Artificial Humanity**

 We need an algorithm for investor behavior so we can counterbalance our least productive actions (e.g., loss aversion, overconfidence, overreaction, etc.)







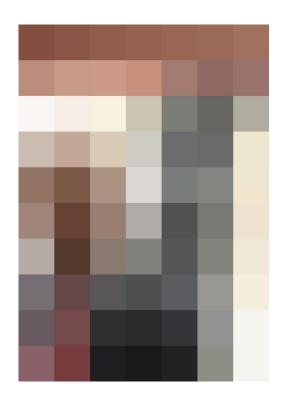


- Expert systems vs. machine-learning techniques
- Expensive storage ⇒ small data, complex code
- Cheap storage ⇒ big data, simple code
- This is closer to natural intelligence! Narrative vs. facts

## Friend or Foe?



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#### Friend or Foe?



<u>José</u>

Susan

- Gender and sex orientation (4)
- Race/ethnicity (4)
- Age (4)
- Current home state (50)
- Religious affiliation (4)
- Political party (3)
- Economic status (3)
- Education (3)

345,600 Possible Types!

But Beware of Learning With Sparse Data

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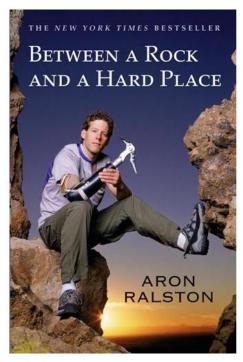




## **Evolution at the Speed of Thought**



Aron Lee Ralston, 4/26/03



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## **Evolution at the Speed of Thought**

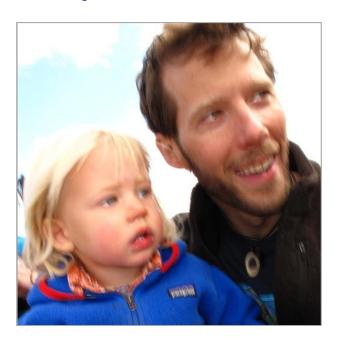
A blond three-year-old boy in a red polo shirt comes running across a sunlit hardwood floor in what I somehow know is my future home. By the same intuitive perception, I know the boy is my own. I bend to scoop him into my left arm, using my handless right arm to balance him, and we laugh together as I swing him up to my shoulder... Then, with a shock, the vision blinks out. I'm back in the canyon, echoes of his joyful sounds resonating in my mind, creating a subconscious reassurance that somehow I will survive this entrapment. Despite having already come to accept that I will die where I stand before help arrives, now I believe I will live.

That belief, that boy, changes everything for me.

Aron Lee Ralston (2005)

## **Evolution at the Speed of Thought**





We Need New Narratives In Finance!

#### Conclusion



#### "It Takes A Theory To Beat A Theory"

- Standard paradigm is not wrong, just incomplete
- Human behavior has been stable for 60,000 years
- Our environment has changed rapidly
- The mismatch can create challenges
- Evolution determines dynamics
- Competition, selection, innovation

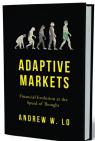
#### How Adaptive Are You?



## Thank You!

## For more on Adaptive Markets:

- http://bit.ly/2t3Sre6 (MIT Sloan Lecture)
- http://bit.ly/2ty6Rqp (Clarendon Lectures)
- http://alo.mit.edu (website)
- @AndrewWLo (Twitter)



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