APAC Regional Webinar: Insights into Quantitative Investing in an Evolving World



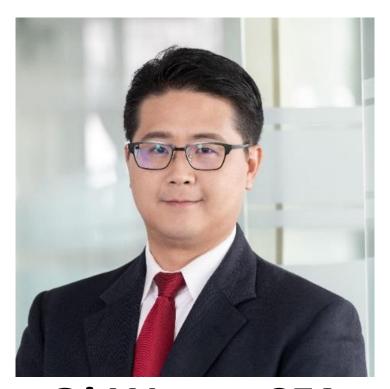
Luis Seco
CEO, Sigma Analysis & Management

23 September 2020, 6:30-7:30 PM HKT

Moderated by Qi Wang, CFA Founder and Managing Director, MegaTrust Investments



SESSION CHAIR



Qi Wang, CFA
Founder and Managing Director,
MegaTrust Investments



HOUSEKEEPING

- Today's webinar is scheduled for 60 minutes
- All participants are muted, we welcome questions via the Q&A function on your screen
- Audience can use CHAT to share comments and to see what others are saying
- This webinar will be recorded and be available to view after the presentation concludes
- At the end of the webinar, please take a moment to complete the short survey. Your feedback is valuable to us.

WE ACKNOWLEDGE THE FOLLOWING AS PRINCIPAL SUPPORTING ORGANIZATIONS:

CFA Society India

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CFA Society Singapore CFA Societies Australia



SPEAKER



Luis Seco CEO, Sigma Analysis & Management









Postmodernism and **Investment Innovation**

Luis A. Seco

GGSJ Center of Digital Management and Technology Innovation



Professor of Mathematics

Director, Mathematical Finance Program



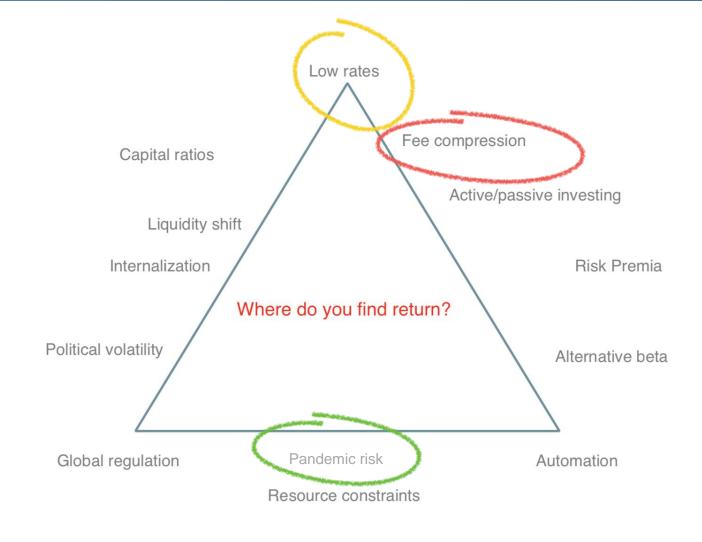








The investment conundrum

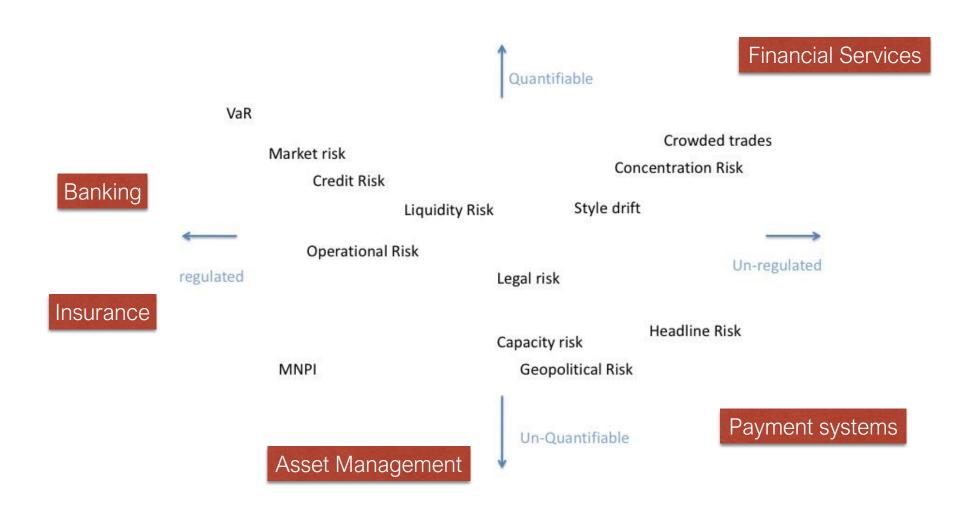








Risk Management









Profitability

Financial Services

Banking

Insurance

Uberization



Googlification

Amazonification

Payment systems

Asset Management

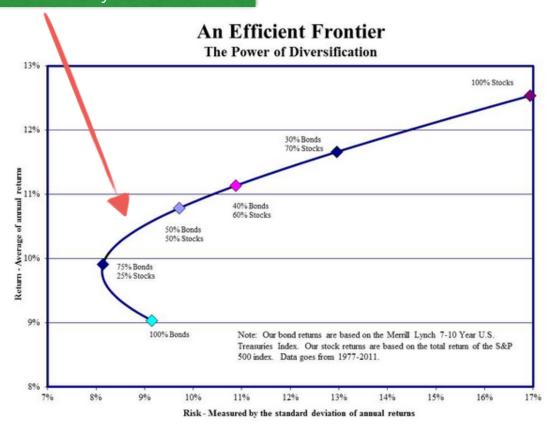






Data Compression 1950 style

Portfolios are represented by two numbers









Modern vs. Post-Modern

Equal weights	$w_i = \frac{1}{N}$
Risk Parity	$oldsymbol{w_i} oldsymbol{\sim} rac{1}{\sigma_i}$
Minimum Variance	$w_i \sim \frac{1}{\sigma_{\partial w_i}^2} = \frac{\partial \sigma}{\partial w_j}$







Al in Portfolio Management

Choice of Methodology (2013 style)

- During Market Crises:
 - Minimum Variance

- During normal markets:
 - 1/N

Market Crises and the 1/N Asset-Allocation Strategy

Marcos Escobar¹, Michael Mitterreiter², David Saunders³, Luis Seco⁴, Rudi Zagst⁵

Abstract

We consider portfolio management strategies where the investment style switches based on the value of a crisis indicator. A variety of strategies is considered in historical backtests on different datasets. Our findings show that certain simple switching strategies achieve statistically significant out-performance when compared to the equally-weighted portfolio with respect to the Sharpe ratio and Omega. In our backtest, the 1/N strategy and equal-risk contribution portfolio perform best during "normal times". On the other hand, during turbulent times, risk considerations seem to play a major role leading to minimum variance as the preferred strategy.

Keywords: Portfolio Allocation, Markov Switching, Equally Weighted, Equal-Risk Contribution

1. Introduction

Since the seminal work of Markowitz (1952), an enormous amount of research has been devoted to the problem of optimal portfolio selection. Recent contributions include the work of Maillard et al. (2010), who study Equal-Risk Contribution (ERC) portfolios, for which the individual risk contribution of each asset to the total portfolio risk is the same, and show that the standard deviations of these portfolios lie







Al in Portfolio Management

We need to do digital to stay relevant

BitCoin is bad BlockChain is good

Automation and digitalization is unstoppable

Automation-inspired job cuts are taboo

Machine learning is a trade-off between a model I can understand and another which may have better coefficients

Yes, overfitting is a big problem, but it is useful for fraud management

Market Crises and the 1/N Asset-Allocation Strategy

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Consulting companies want automated nirvanas devoid of people, except for their consultants

ching, Equally Weighted, Equal-Risk

ince as the preferred strategy.

Data science is exciting yet dangerous, profitable for some, a time-sink for others

unt of research it contributions ribution (ERC)

portionos, for which the individual risk contribution of each asset to the total portains risk is the same, and show that the standard deviations of these portfolios lie

you say machine learning; I say overfitting







Empirical Analysis in Portfolio Management

Choice of Methodology (2020 style)

- Define Market conditions
 - · Let the machine learn

The Machine Rules:

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Machines learn from humans









Post-modern Portfolio Theory

Agnosticism

Social Science

Negationism

Nicholas Nassim Taleb: "Your Grandmother's Intuition about Tail Risk is Usually Right"

HSBC's Head of Global Risk Strategy:

"it is important to have historians, geographers and political scientists alongside the mathematicians in formulating scenarios."

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The Basel Committee has proposed retraction of its AMA to operational risk, preferring "Standardized Measurement Approach" (SMA).

Was AMA too deterministic and hence we seek to reduce the focus of analysis?

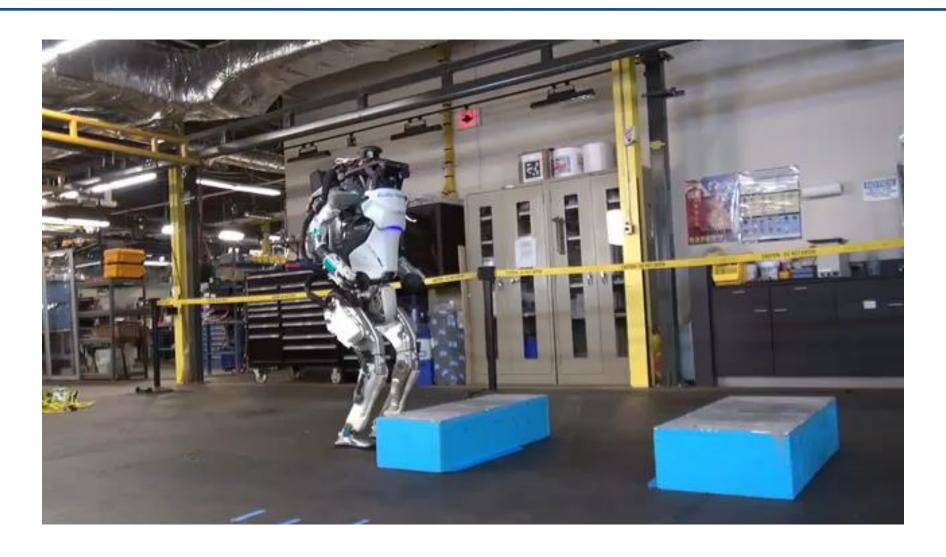
Does this allow expanding operational risk to incorporate compliance, fraud, cyber, etc?.







Machines fail just like humans



2014: The Calpers Syndrome

- In 2014, the California Public Employees
 Retirement System stunned the investment
 world with the announcement that they
 were exiting their multi-billion dollar hedge
 fund portfolios
- While the fund had numerous problems, fees paid to hedge funds was announced to be one reason.



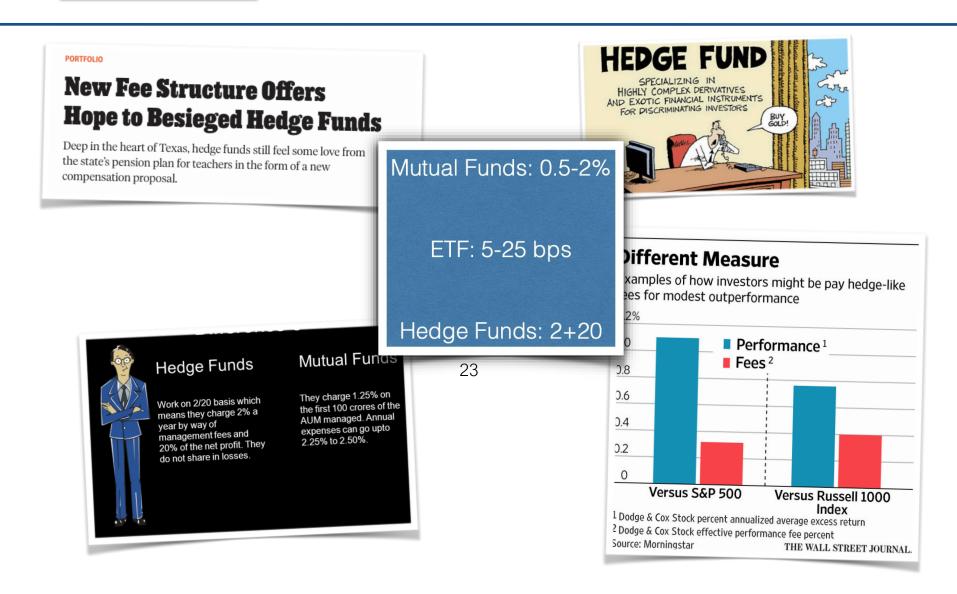






Profitability is at risk

The Talk on Fees









Optimal Fees

- The Fund manager will build a portfolio to optimize <u>its own utility</u>: the fee income
- The investor can -indirectly- select the fee it pays by choosing to participate in the fund or not
- The rational investor will aim for a fee that will optimize its own utility: the return after fees

Fee Innovation

Fund managers are offering performance protection in cash....

...no management fees...

... in exchange for higher performance fees.



Negative fees?

	Investor	Manager
Invested amount	2000	200
Gross investment gains	100	
Guarantee	0	0
Performance fee	-50	50
Total gains	50	50
Profitability	2.5%	25%

Investor/ manager roles are reversed

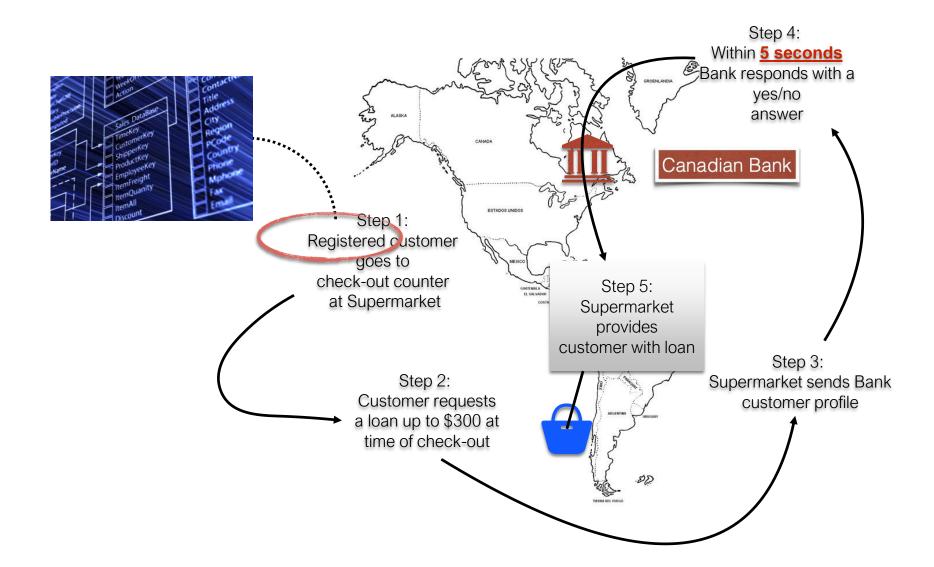
	Investor	Manager
Invested amount	2000	200
Gross investment gains	-160	
Guarantee	0	-160
Performance fee	0	0
Total gains	0	-160
Profitability	0%	-80%







Post-modern banking









Big data - Big Risk

Castonici mann

Address Profession Income Address

Shopping patterns by

- time
- goods purchased
- store locations

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Customer Database



Subject to manipulation

Dynamic database: updates in real time

Decision and monitoring systems updated in real time

Not Bank-owned

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Machine Friendly Human un-friendly

Unstructured Data

Customer name vs Household name Customer address vs. store address

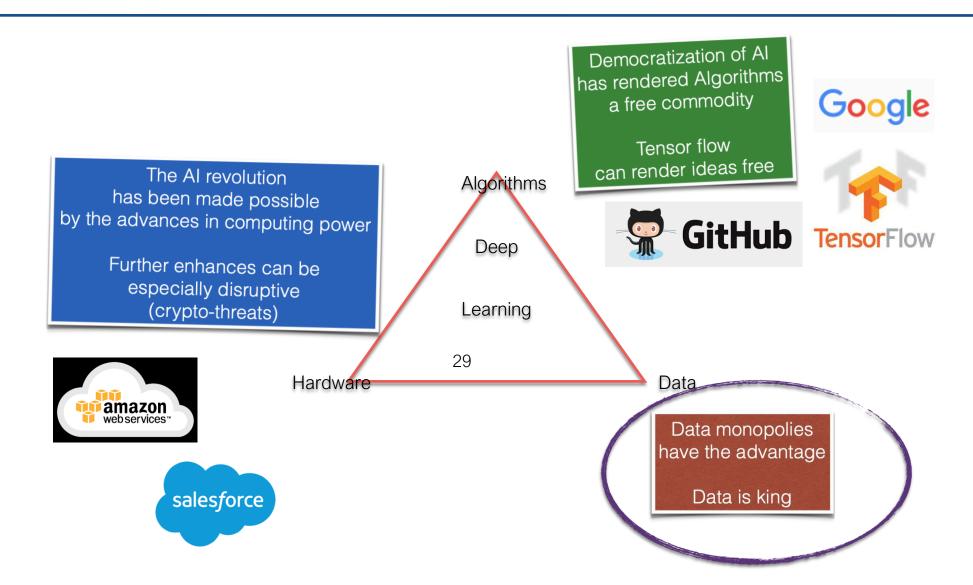
Mining the customer network







Al at work









Scotiabank and Deep Learning

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Dozens Recount Pattern of Sexual Misconduct by Vegas

Deaths of Toronto Billionaire and Wife Now a Homicide Case

Trump Promotes



Elliott Management Discloses Stake in Sky



Search Q

CIO JOURNAL

Scotiabank Deploys Deep Learning to Improve Credit Card Collections

Tool helps identify potentially delinquent customers and suggests ways to approach them

By Steven Norton

Feb 6, 2017 5:35 pm ET

O COMMENTS

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Deloitte

Hashiam Kadhim

the best way to approach them about it.



DeepLearni.ng Lead Machine Learning Engineer



Michael Zerbs, chief technology officer at Scotiabank, hopes to deploy machine learning across the bank to better understand customers. PHOTO: BANK OF NOVA SCOTIA

Even a small improvement in collections can have a big impact on the bank's bottom line, said Michael Zerbs, chief technology officer at Scotiabank. The challenge is to figure out the best way to interact with customers who may have simply forgotten to pay versus those less likely to pay at all.

For customers who just forgot, "you actually don't want to call them because it's a bad customer experience" and wastes bank resources, Mr. Zerbs said. A simple nudge via text message might be simpler and more

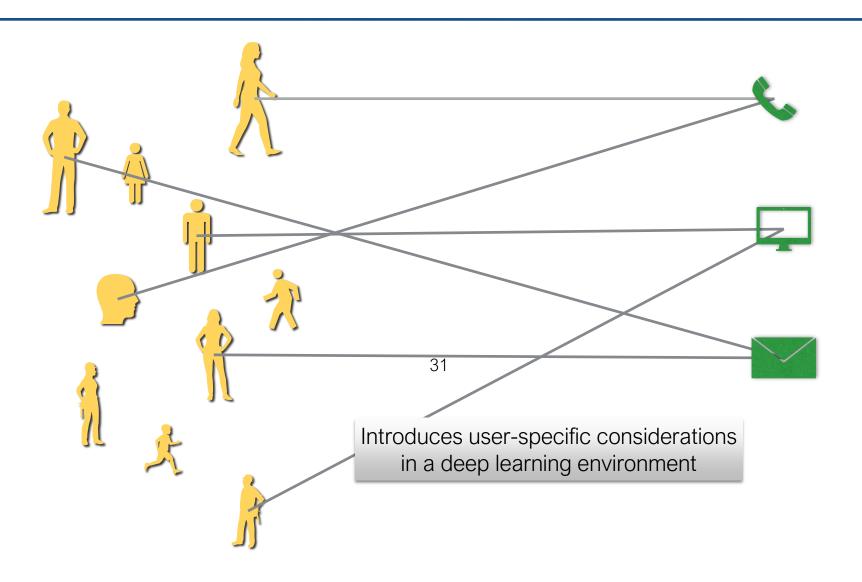
effective.







Collections: End-to-end solutions









Investing in the time of Populism

ESG - Credit Scoring based on

- Environment
- Social
- Governance

The Amazon-ification of Asset Management



Cryptocurrencies

Portfolios that sell







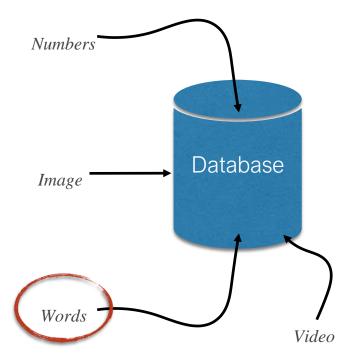
Data

Traditional data sources are being replaced by unstructured data sources

- Facebook, twitter, news outlets...
- Sensors
- · Cloud-resident data

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Algorithms will have to adapt









Application: AI-ESG

Construct ESG scores using web-resident, text data sources

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- Company document centres
- Sentiment analysis sourced in real time
- News sources

Building Machine Learning Systems to Automate ESG Index Construction

> May 19, 2020 Alik Sokolov, Jonathan Mostovoy, Jack Ding, Luis Seco







Regulation

1929 Crash	Securities ACT (1933)
Enron	SOX (2002)
2008	Dodd-Frank (2010)
The second secon	Data Regulation?
COVID-19	?







The new financial players

1970's	Private banking/Hedge Funds
1990	Derivatives/Prop desks
2000	³⁶ Pensions/Asset owners
2020	Family Offices

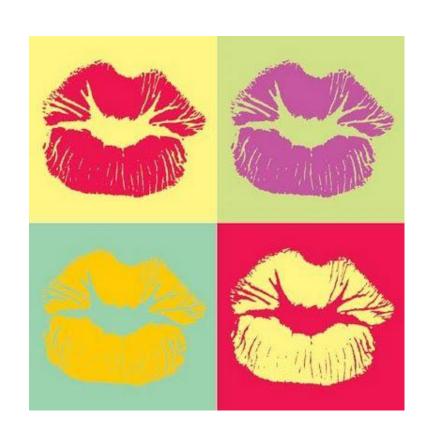






The end

谢谢



Thanks