



Advisors are under immense pressure...



A new type of client

- · Investors questioning value of advice, level of services, user experience, transparency, fees
- Change in demographics greater emphasis on real-time information, accessed through multiple platforms (mobile, web)

Regulation

• Tighter regulatory environment placing much higher demands on client interactions, disclosure and transparency

Fee compression

Downward pressure from clients, increased competition, low-fee products, robo-advisors

Complex markets

Looming correction, lower returns & alpha, talks of recession...



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Technology Gap & Operational Inefficiencies



Wealth management firms have been underserved by technology

Compared to institutional money & banking

2000s Buy Side

- Asset managers
 & hedge funds
- Risk & portfolio analytics, trading strategies

2015+ Wealth Management

- Goal-based investment & portfolio management
- Risk & performance
- Communications: advisors & individual investors

1990s Banks

- Risk management & trading platforms
- Banking regulation

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And the Pressure is Mounting ... Morgan Stanley launches new

advisory technology platform





KEY POINTS

Morgan Stanley 's wealth-management leaders won't say exactly how much the business is spending on technology, but they want to make clear that it's Morgan Stanley has launched a reserving as one destination for all a lot. Think a "big number," with "lots of commas,"



The online advisor platform, WealthDesk, brings to a single dashboard a consolidated view of client relationships and portfolios... it integrates portfolio advice, portfolio risk measurement, fee and pricing information and consolidated view of clients' plans and portfolios updated each quarter...

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A New Advisory Model



The new advisory model based on

a transparent value add practice which redefines and builds long-term trust...

Digitally driven

• End-to-end workflow and enhanced client communication

Client-focused automated business processes

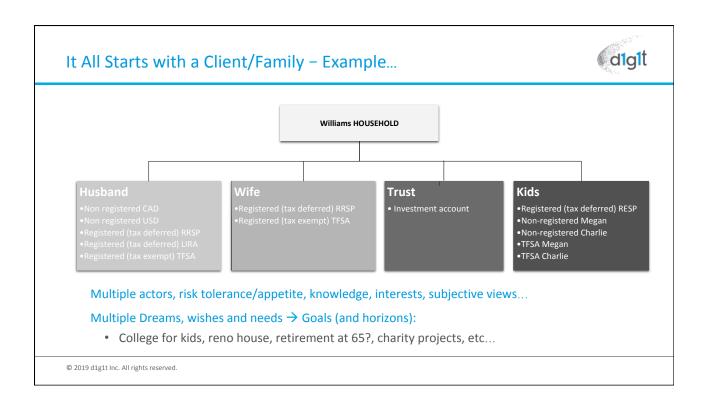
PM, compliance, and communication defined by client's goals and portfolios... not the firm's products

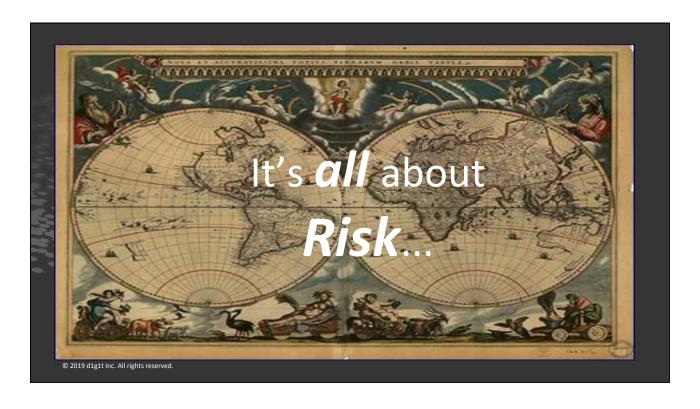
Investment sophistication

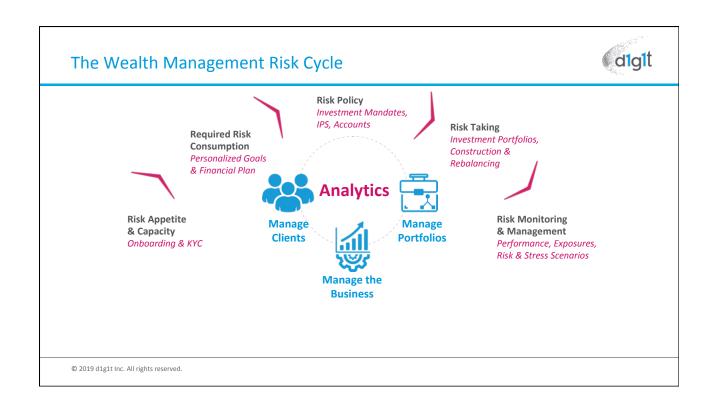
Technology-driven investment process guided by analytics and rigorous risk management

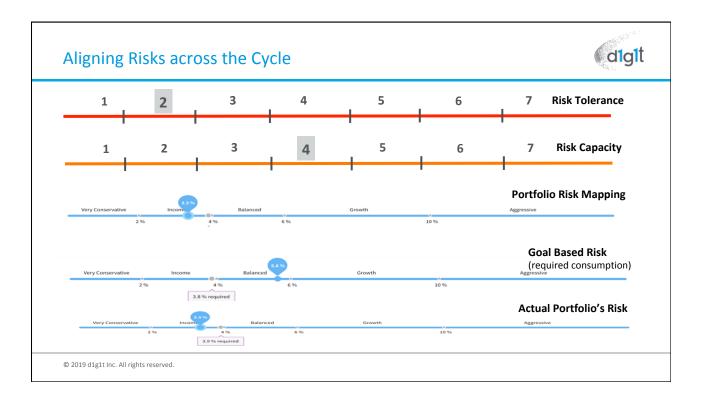
High-touch services

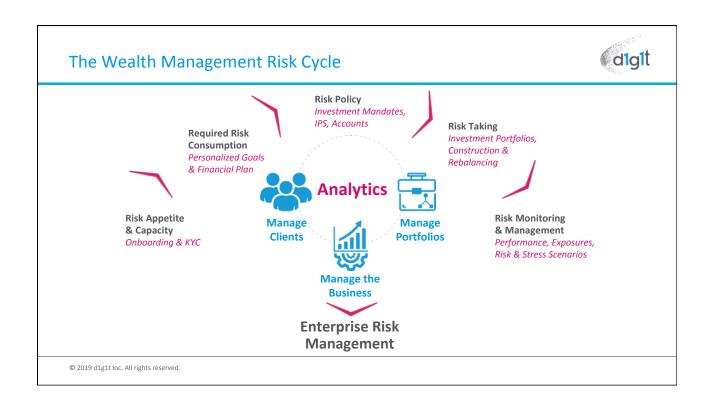
Personalized high-value, interactive client engagement, increased communication and trust

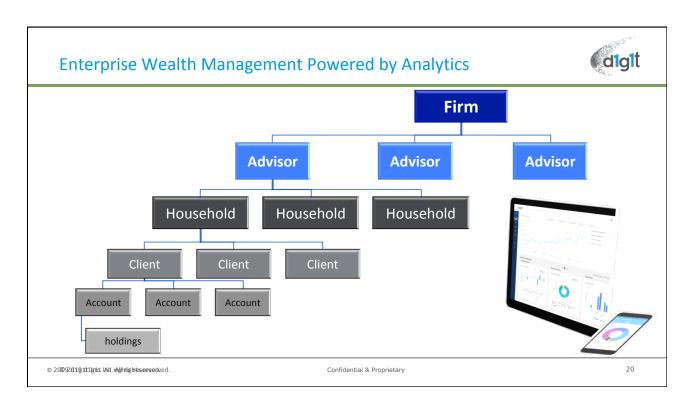












Analytics to Manage The Wealth Management Risk Cycle



The modern Wealth Management analytics toolkit integrates

- Data Science / Machine Learning
- Behavioural finance
- Goal Based financial Planning
- Modern Portfolio Theory, Portfolio Optimization
- Risk management tools
- ... Long-term investment focus

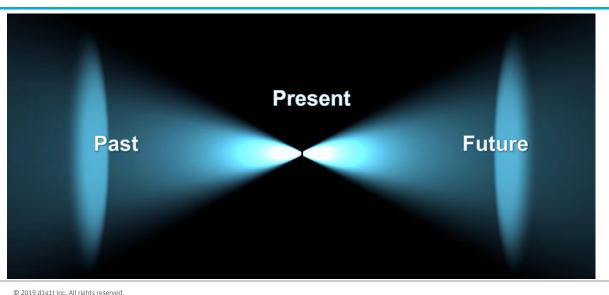


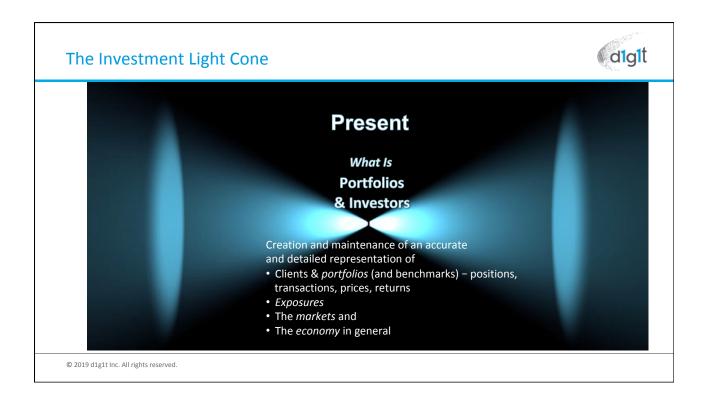
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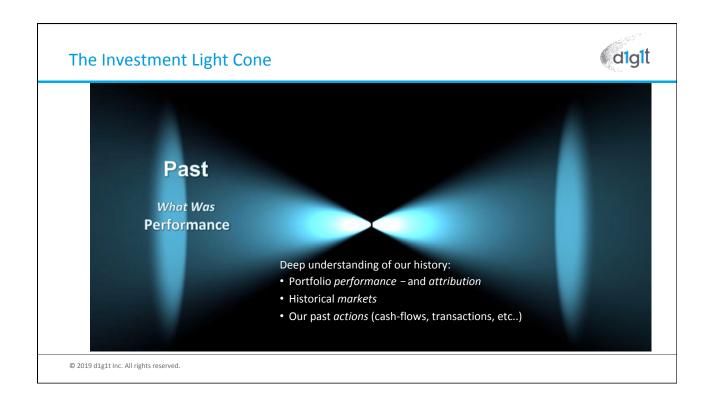
Analytics – The Investment Light Cone

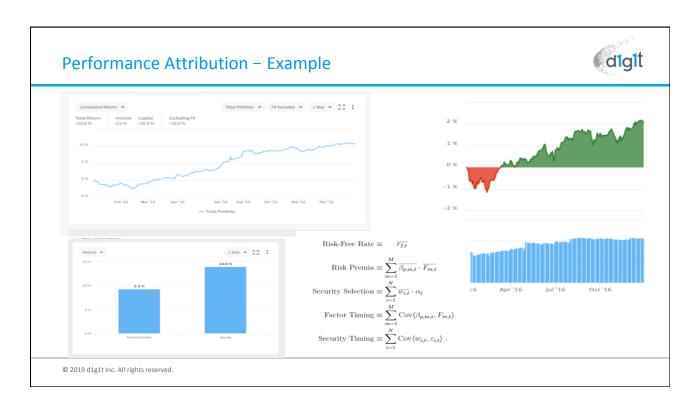






€d1g1t Portfolio Exposures - Example \$ USD (Million) NMV NMV Short USD EUR GBP JPY Total Long Portfolio 457.5 567.8 110.3 EQ 32% 4% 11% 5% 52% EQ 221.4 225.5 4.1 IR 10% 14% 24% IR 0.0 111.2 111.2 CR 18% 7% 24% CR 124.9 231.1 106.2 Total 59% 32% 5% 100% Portfolio 458 **US Equities** 221 Utilities. 1% Consumer Disc, 7% Telecom, 11% USD-EQ 136 EUR-EQ 48 Information Technology, 10% GBP-EQ 16 JPY-EQ 22 IR 111 USD-GOV 52 EUR-SOV 59 125 USD-IG 70 USD-HY 60 **EUR-HY** 32 Source: S&P Capital IQ. For illustrative purposes only. USD CDS -38 © 2019 d1g1t Inc. All rights reserved.





Learning from History - Example: Regression Engine



Objectives:

- · Provide on-the-fly knowledge and attribution (what, how, & why) about client engagement
- Understand overall Enterprise Business Drivers
- Test and construct strategies

ML Engine is a combination of:

- Data generation and normalization: return computation and financial modeling (multiple sources)
 - Detailed client information (over time)
 - Client portfolios: multiple portfolios, hierarchies and holdings level information
 - Positions, transactions, prices detailed feature generation and segmentation: e.g. currency, capital gains, realized-unrealized, taxes, dividends and cash-flows, ...)
 - Market information (external portfolio/benchmarks), as well as events/news etc.
 - Factor analysis and economic indicators
- Econometrics and Machine Learning toolkit (unsupervised learning, regression, NNs, RL)

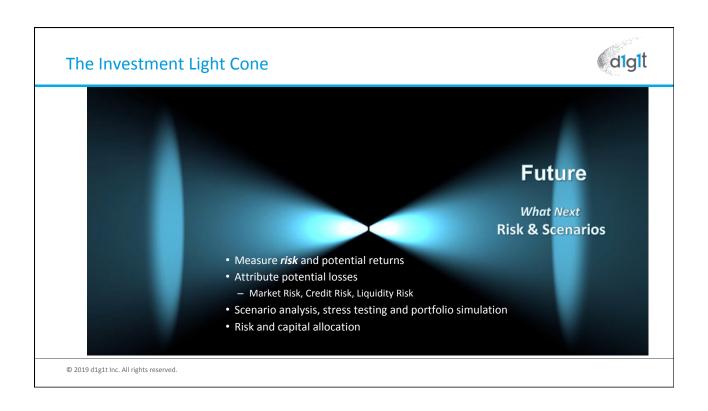
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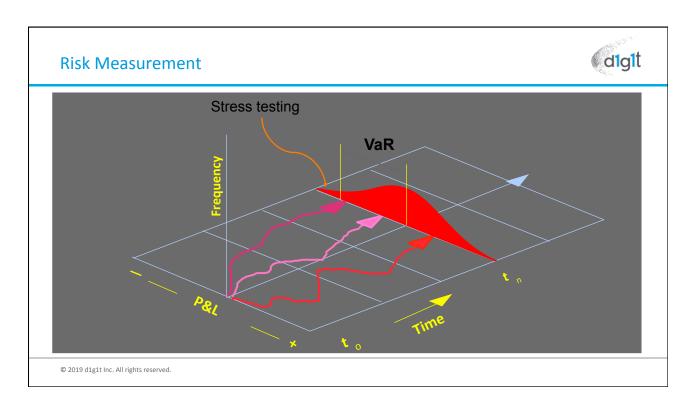
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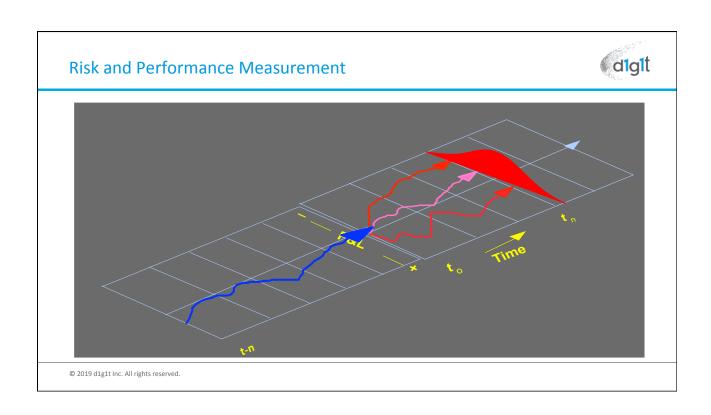
The Investment Light Cone

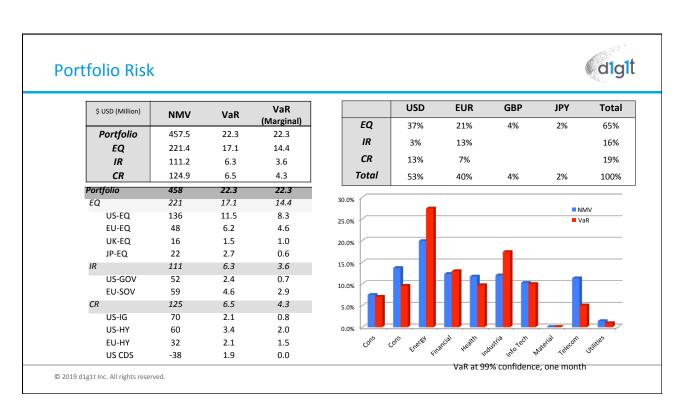












The Investment Light Cone What Is Portfolio What Was Performance What Next Risk & Scenarios Design and testing of strategies, reactions and intervention measures (management, policy modelling, portfolio construction, governance)

Example: Linking Risk & Goal Based Wealth Management

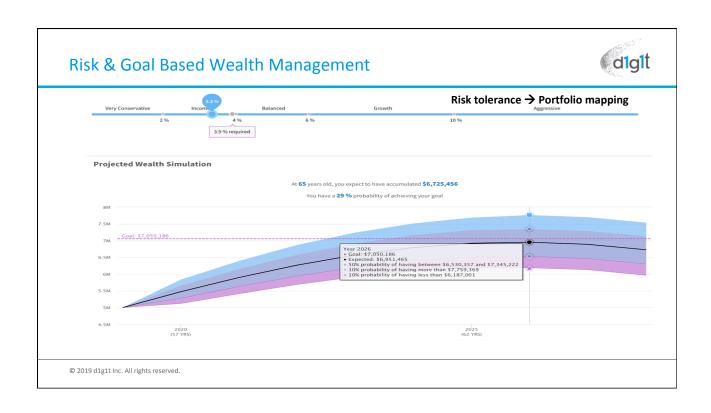


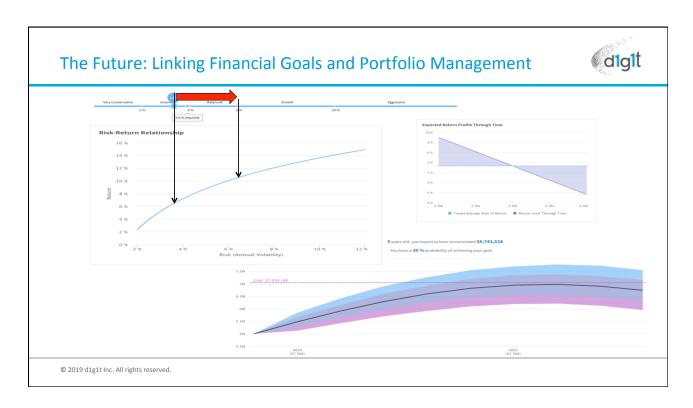


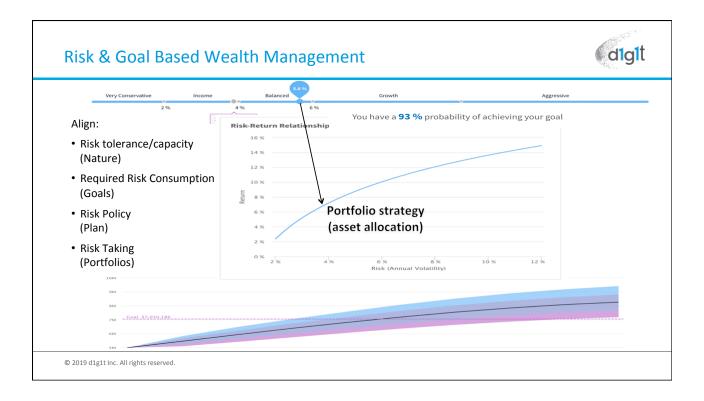
Key points:

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- We can consistently combine:
 - · Goal Based financial Planning
 - Modern Portfolio Theory, Portfolio Optimization tools
 - Behavioural finance
- Focus on (long-term) risk measures related to
 - Probability of achieving Goal(s)
 - Distance to achieve goal(s)
- Create explicit link to portfolio risk measures (volatility, VaR, drawdown, etc...)







Goal base Wealth Management



Goal Based Portfolio Optimization – intuitive and actionable for financial advisors

- Maximize the likelihood of achieving the goal (or set of goals)
- Minimize a distance function (e.g. mean squared errors) to achieving goal

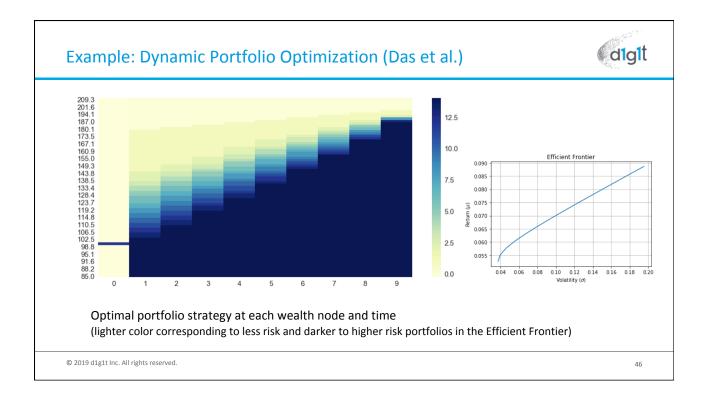
Key insight: integrate MPT by choosing portfolio set on the efficient portfolio (splitting optimal asset allocation and the required risk taking to achieve a goal)

Example - Das et al

- Static (constant risk) solution
- Dynamic goal based portfolios (using stochastic dynamic programming)
- Trading off Multiple goals
- Dynamic portfolios via Reinforcement Learning

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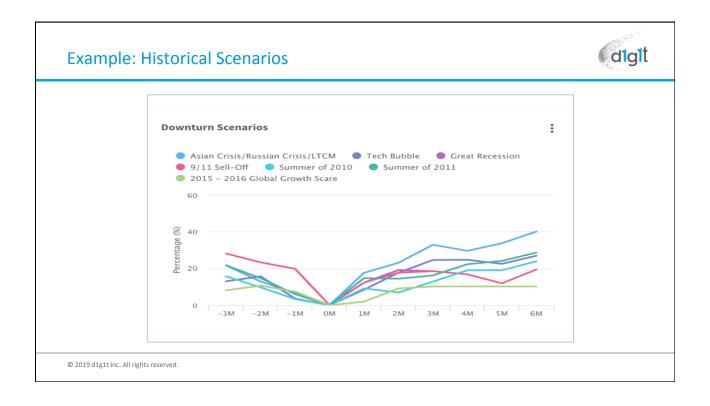
The Investment Light Cone – Scenarios





The quality of risk analysis and the actions/decisions relies on our ability to generate **relevant, plausible** and **comprehensive, forward-looking scenarios** that properly represent the future





Scenario Analytics



The development of relevant forward-looking scenarios requires the combination of

- Detailed understanding of portfolios, clients and goals
- Good economic analysis
- Grounded financial models
- Data science and econometrics

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Example: Scenarios from Economic Report



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The growth prospects for the U.S. economy remain fairly favorable, had restrained to positive momen half of last year.

40 Cheap Dil And An Expansive QE Program Under The Eurozone Recovery By Jean-Michel Six, Paris The financial news, expected start t for full-year GDI

The Eurozone Recountry

By Jean-Michel Six, Paris

The financial news, and stock and bond markets, have bershort of since the the year may son excessive complete irrationa the real reached fastest of the process differed from one country to the next, with various combinations or integrating into the global economy and moving on the value chain in goods

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of integrating into the global economy and moving up the value chain in goods and services, state-led urbanization, and the increased urbanization and open of a rising middle class.

Economic Outlook – Global Scenario

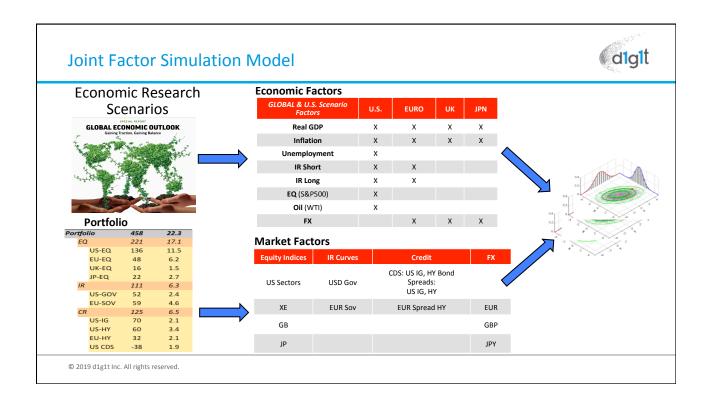




GLOBAL ECONOMIC OUTLOOKGaining Traction, Gaining Balance

GLOBAL Scenario 2015	US	EURO	UK	JPN
Real GDP (% change)	3	1.5	2.8	0.8
Inflation (% change)	-0.3	-0.3	0.1	0.4
IR Short	0.3	0.1	0.5	0.05
IR Long	2.3	0.3	2.0	-
FX		0.9	0.7	120.0

Source: Standard & Poor's Ratings Services economic research report dated April 22 2015. Indexes are unmanaged, statistical composites and it is not possible to invest directly in an index. These results are inherently limited because they do not represent the results of actual trading and were constructed with the benefit of indiscipit. The returns shown do not reflect payment of any sales charges or fees an investor would pay to purchase the securities they represent. The imposition of these fees and charges would cause actual and back tested performance to be lower than the performan shown. Returns exclude divisioning.



Example: Joint Factor Simulation Model



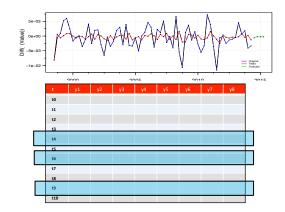
Input: quarterly data for all the factors (20+ years)

1. Marginal processes for each factor: ARMA GARCH model (filtering)

$$y_t = a_0 + a_1 y_{t-1} + b_1 \epsilon_{t-1} + \epsilon_t$$

$$\epsilon_t = \sigma \eta_t, \quad \sigma_t^2 = \alpha_0 + \alpha_1 \epsilon_{t-1}^2 + \beta_1 \sigma_{t-1}^2.$$

2. Historical codependence of residuals (allows for non-Gaussian fat tails and tail dependence)



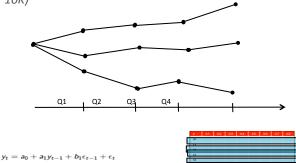
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Joint Factor Simulation Model

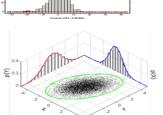


Empirical (simulated) factor terminal distribution (horizon = 1 year)

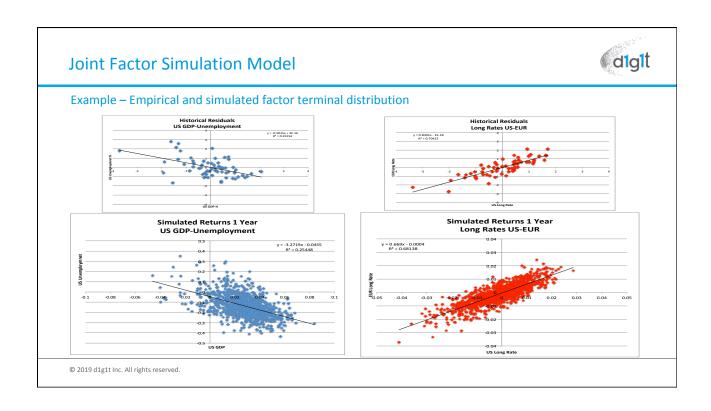
Monte Carlo simulation of N scenarios of the joint factor processes over 4 quarterly steps (N = 1K - 10K)

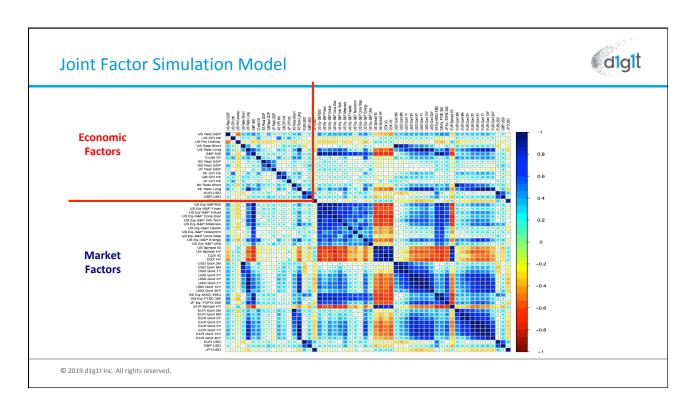






 $= \sigma \eta_t$, $\sigma_t^2 = \alpha_0 + \alpha_1 \epsilon_{t-1}^2 + \beta_1 \sigma_{t-1}^2$,



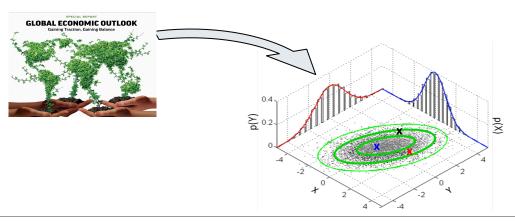


Economic Scenarios – Model Mapping



Objective: express forecasted scenarios in the context of joint factor simulated distribution

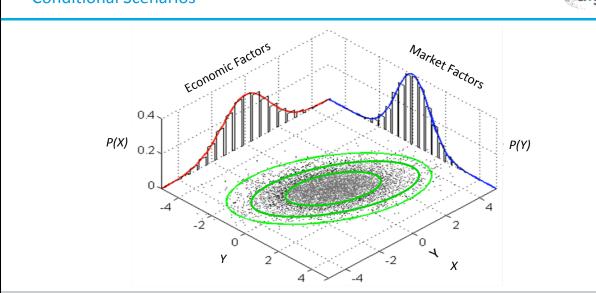
 Standardized economic scenarios (expressed as number of standard deviations of factors, and also in terms of likelihood within the model)

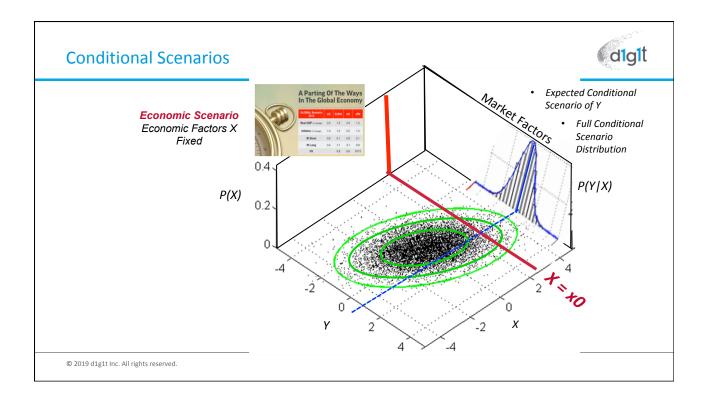


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Conditional Scenarios







Conditional Scenarios: Analytical Methods



Conditional factor distributions are available analytically for certain joint distributions e.g. Multi-variate Gaussian

$$X = \begin{pmatrix} X^{(1)} \\ X^{(2)} \end{pmatrix} \sim N(\mu, \Sigma)$$

$$\mu = \begin{pmatrix} \mu^{(1)} \\ \mu^{(2)} \end{pmatrix}, \qquad \Sigma = \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix}$$

Conditional distribution of $X^{(2)}$ given $X^{(1)} = x^{(1)}$ is multivariate normal with mean m and covariance matrix B

$$\begin{split} m &= \mu^{(2)} + \Sigma_{21} \Sigma_{11}^{-1} (x^{(1)} - \mu^{(1)}) \\ B &= \Sigma_{22} - \Sigma_{21} \Sigma_{11}^{-1} \Sigma_{12} \end{split}$$

Conditional Scenarios: Least Squares Stress Testing (LSST)



Regress Under Stress

A Simple Least-Squares Method for Integrating Economic Scenarios with Risk Simulations

Dan Rosen, David Saunders 2

Journal of Risk Management in Financial Institutions, 9(4)

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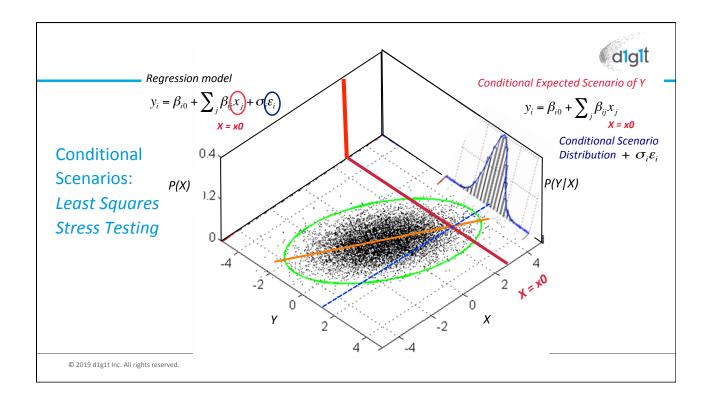
Conditional Scenarios: Least Squares Stress Testing (LSST)

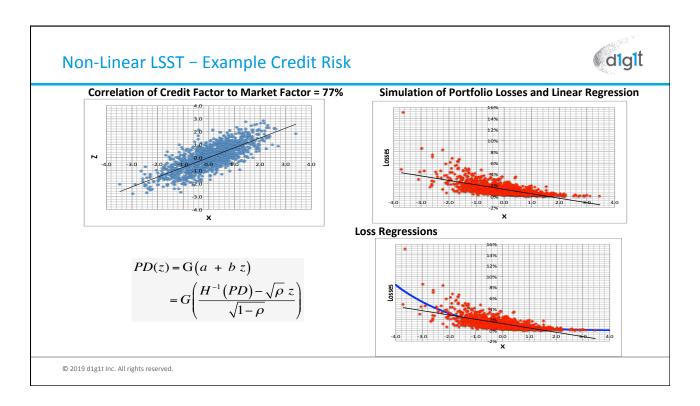


Key insight: conditional expectation of all the factors (and more generally the full conditional distribution) can be estimated directly from a pre-computed simulation using Least Squares Regression (LSR) or more sophisticated ML methods

- Conditional scenario analytics, including risk factor contributions, can be derived from the regression results
- The application of LSR on the cross-sectional information of a simulation to obtain conditional expectations is the key component of LSM to price American options (Longstaff and Schwartz 2001)
- · Applied here to portfolio risk management and stress testing

Other ML examples: application of Neural Networks to simulate derivatives portfolios, compute greeks, etc...





Risk and Scenario Analytics through Regression (ML) Engine



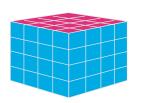
Data generation and normalization: Pre-computed simulation Cubes (expensive step)

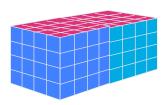
- Holdings- and factor-level simulations from risk engine
- Detailed portfolios: multiple portfolios, hierarchies and holdings level information
 - Positions, transactions and prices, with detailed feature generation and segmentation: e.g. currency, capital gains, realized-unrealized, taxes, dividends and cashflows, ...)
- Enriched Cube: risk factors economic indicators

Full Machine Learning toolkit (unsupervised learning, regression models, reinforcement learning, ...)

Applications: scenario analysis, strategy construction/validation, optimization, goals-based portfolios

Mark to Future





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Portfolio Simulation And Analysis – Global Scenario



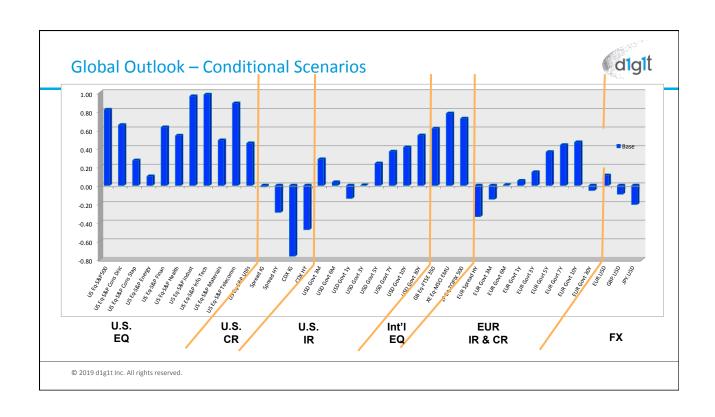
	NMV	VaR (Annual)	P&L Mean	Global Scenario	Rel. Return
Portfolio	457.5	16.0%	3.7%	8.1%	4.4%
EQ	221.4	29.0%	6.7%	17.1%	10.4%
IR	IR 111.2 14.49	14.4%	1.6%	-0.1%	-1.8%
CR	124.9	16.8%	0.4%	-0.5%	-0.8%

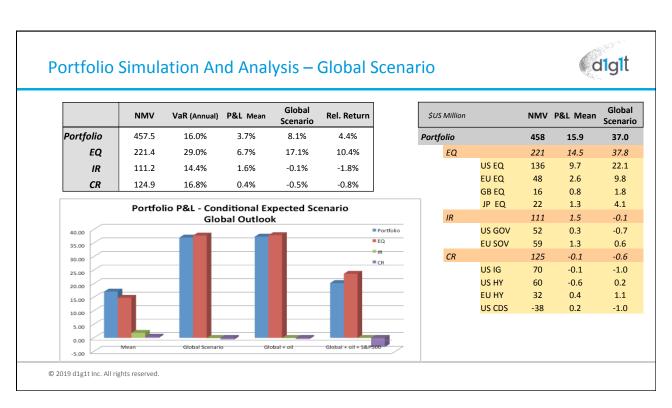


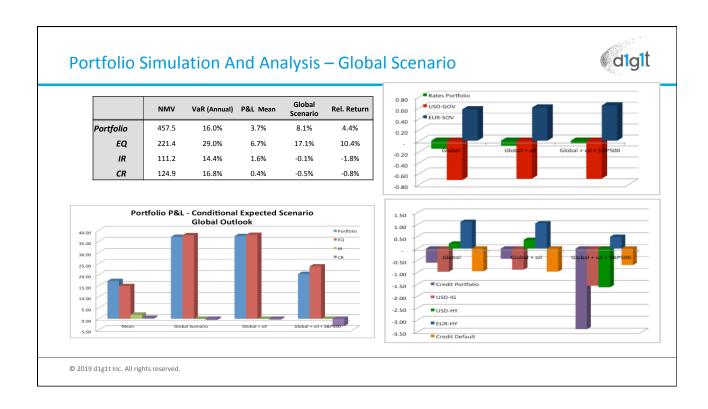
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FX		0.9	0.7	120.0

ŝus	Million		NMV	P&L Mean	Global Scenario
rtfo	olio		458	15.9	37.0
	EQ		221	14.5	37.8
		US EQ	136	9.7	22.1
		EU EQ	48	2.6	9.8
		GB EQ	16	0.8	1.8
		JP EQ	22	1.3	4.1
	IR		111	1.5	-0.1
		US GOV	52	0.3	-0.7
		EU SOV	59	1.3	0.6
	CR		125	-0.1	-0.6
		US IG	70	-0.1	-1.0
		US HY	60	-0.6	0.2
		EU HY	32	0.4	1.1
		US CDS	-38	0.2	-1.0



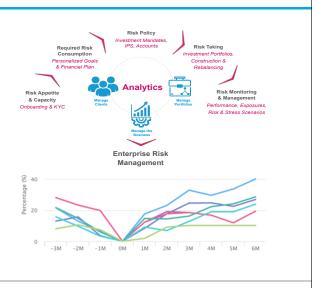




Remarks: Scenario Analysis as Advisory Tool Over the Entire Cycle



- Elicit and define client's risk appetite and tolerance, clarify goals, and design investment/risk policies and IPSs
- Portfolio management
 - Design specific client portfolios
 - · Monitor and attribute risk
- Powerful communication tool to help engage and manage clients
 - On-going client communication around life events, market downfalls, crises, economic opportunities
- Engage prospects: investment and advisory proposals for new clients or new plans





Technology will empower

wealth management firms

to scale up the high-value, human services

that can set them apart

in an increasingly automated & digital world.

