

OLAP for Optimum Asset Allocation under Uncertainty

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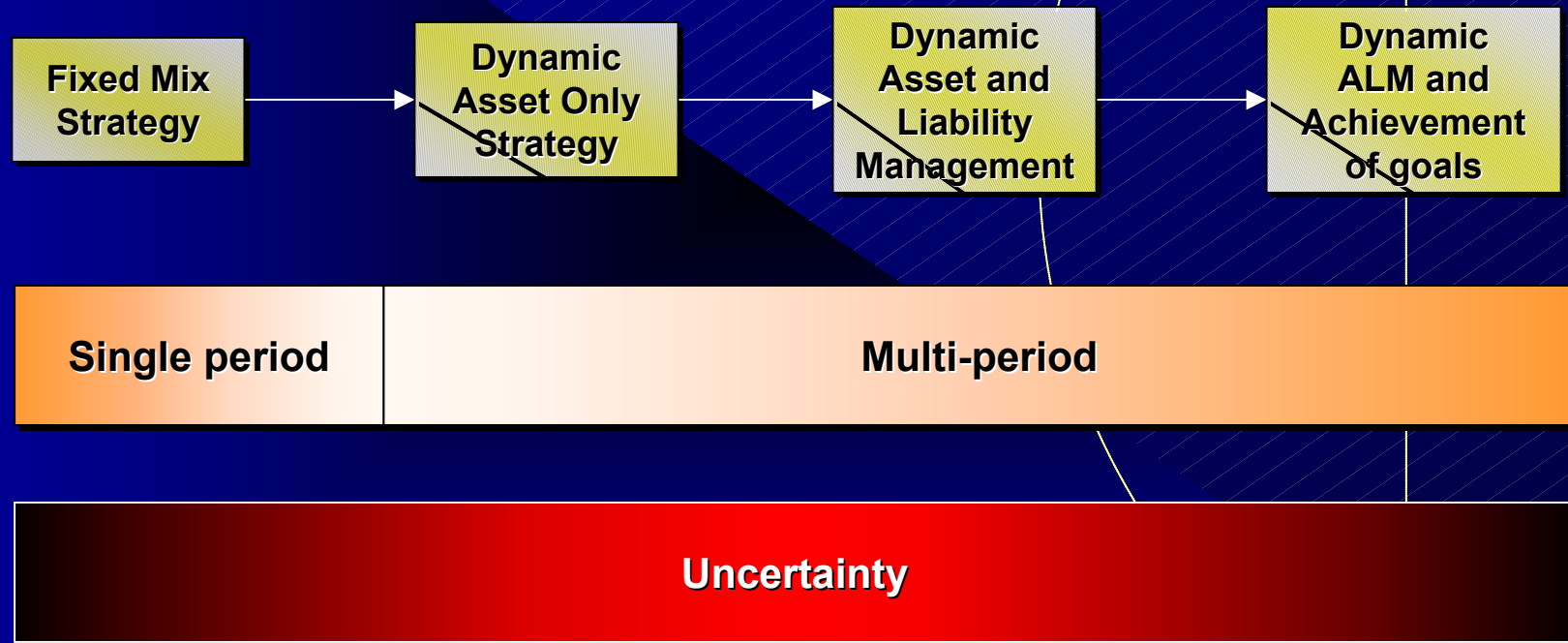
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**Special Focus Symposium on Decision Technology
and Intelligent Information Systems**

Outline

- ◆ **Optimum Asset Allocation**
- ◆ On-Line Analytical Processing for DSS
- ◆ OLAP for Asset Allocation using 2-stage Stochastic Programming

Evolution of Optimum Asset Allocation Models



Disadvantages of traditional approaches (1/2)

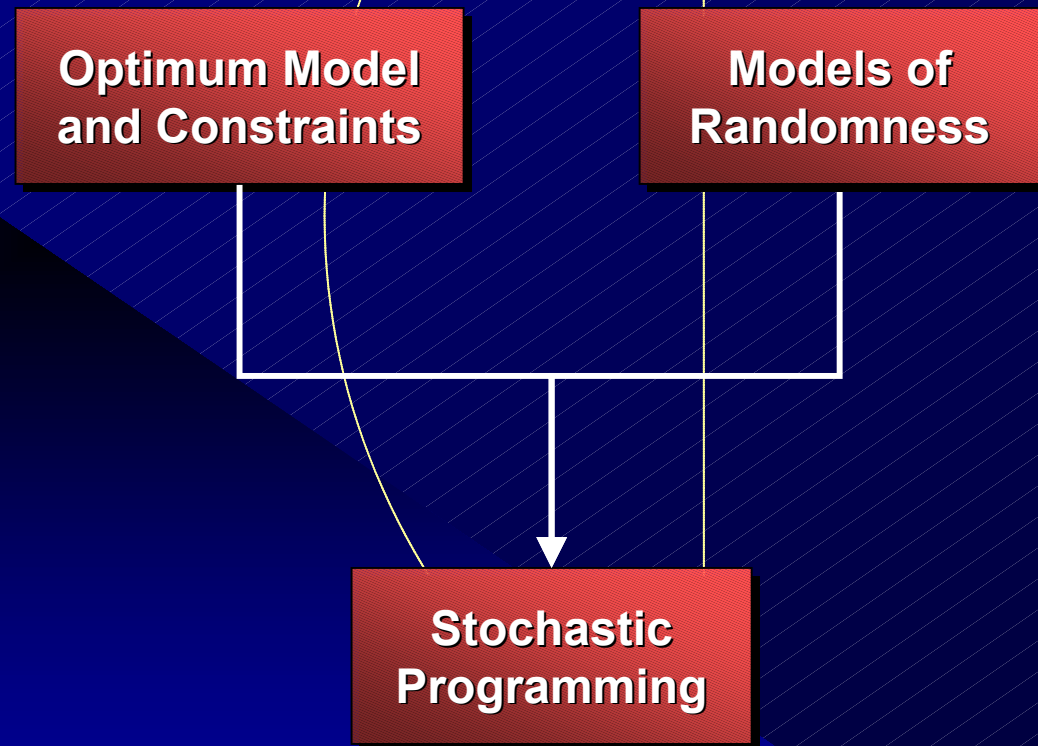
- ◆ **The single period nature of the model.** No long-term impact of short-term objectives.
- ◆ **The effect of errors in mean.** Although errors in variances in comparison with errors in co-variances seem to have twice a damaging impact in the decisions, the effect from errors in means have ten times greater impact.
- ◆ **The symmetric definition of risk.** Such a measure penalises negative and positive returns.
- ◆ **Computational burden.** Markowitz recommends $n(n+1)/2$ constants have to be computed for the model. Large-scale, dense quadratic programs, are difficult to solve for $n > 500$. (MAD formulations overcome this difficulty)

Disadvantages of traditional approaches (2/2)

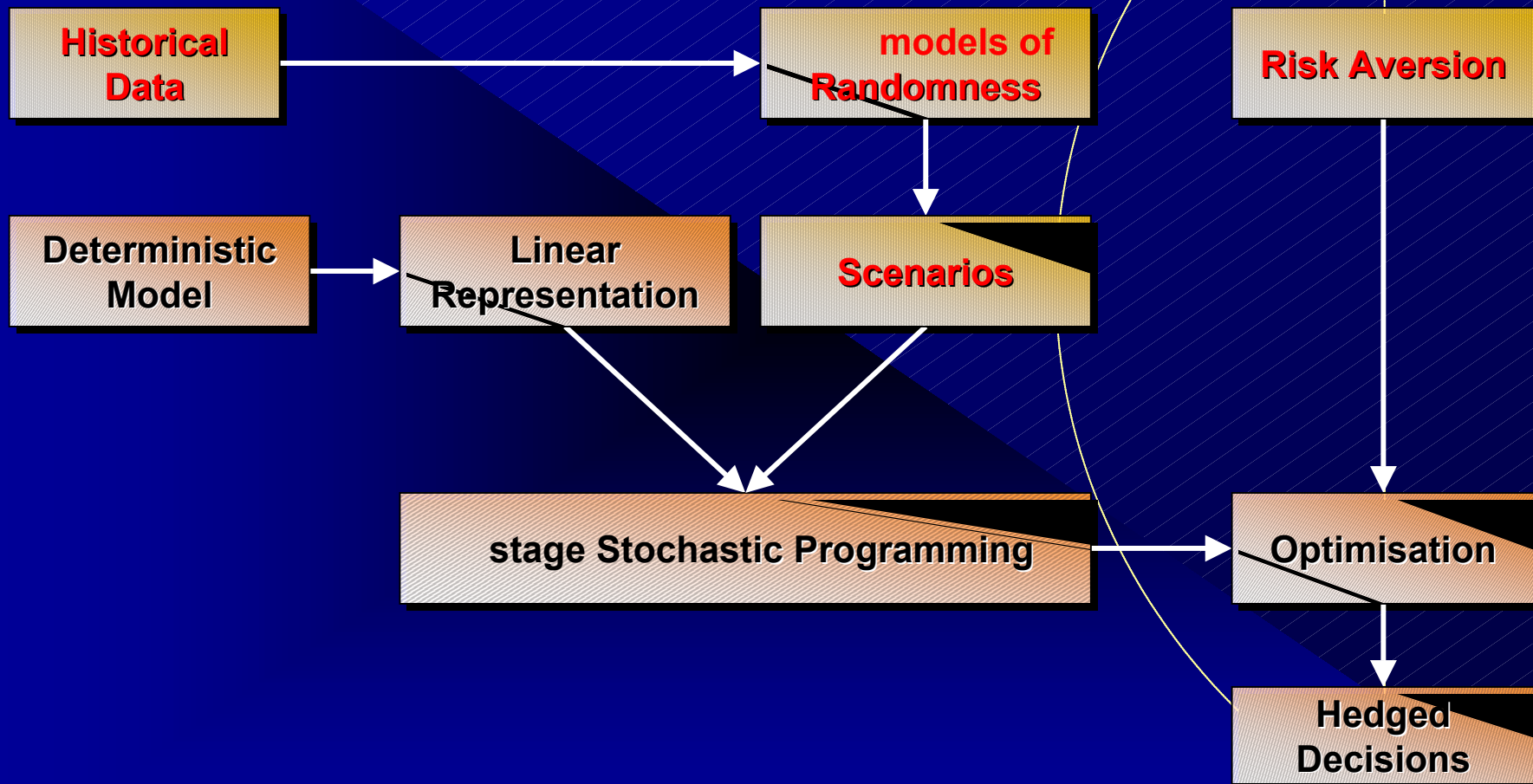
- ◆ **Transaction costs.** An optimal solution often results in many buy and sell decisions which can be costly if transaction costs are taken into consideration.
- ◆ **Management costs and cut-off effect.** Typical optimal solutions usually suggest a portfolio with a large number of assets; a difficult task that is not practised in reality. Also trading decisions may not meet the minimum transaction lots. Either the decisions need adjustment or a more complex integer quadratic program has to be solved.
- ◆ **Randomness $\sim N(\mu, \sigma)$. Randomness is Normally distributed.** The Normal Distribution is not necessarily the 'norm.'

Why Stochastic Programming?

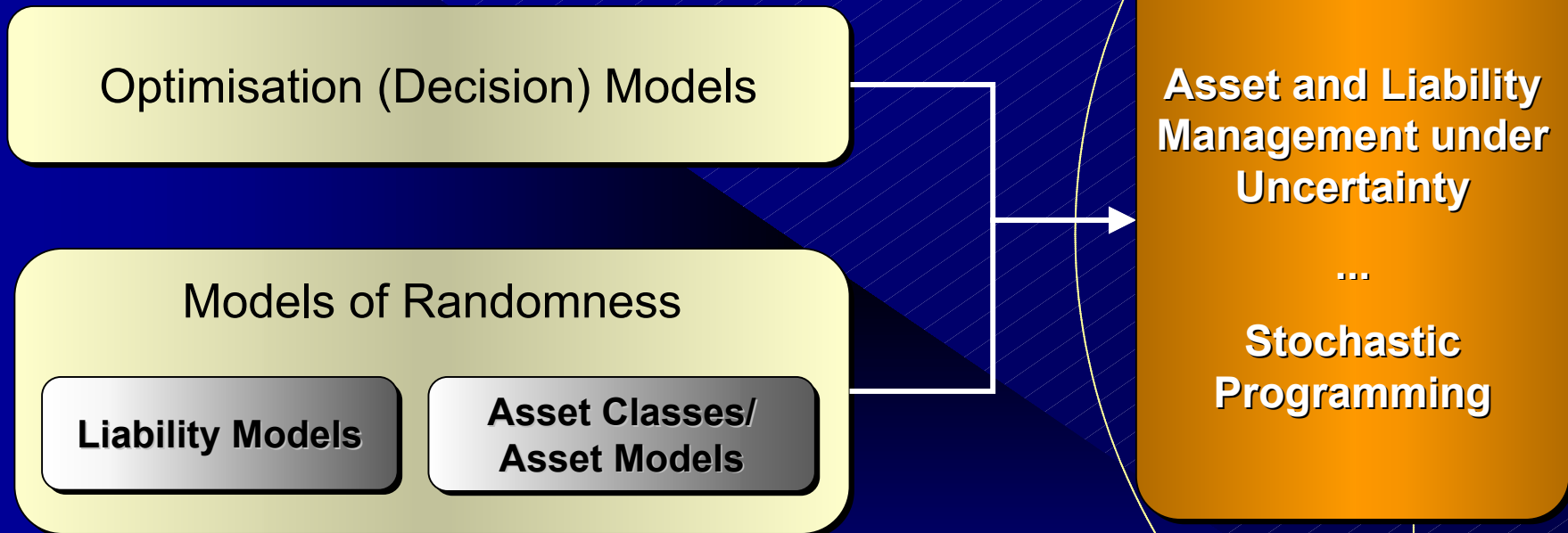
- ◆ Planning for the Future is synonymous to dealing with uncertainty.
- ◆ Optimisation (Linear/Integer Programming) models assume certain states of nature.
- ◆ Multi-time period of investment decisions.
- ◆ Decisions need to be made today, “Here and Now.”



Stochastic Programming



ALM and SP Integration



Integrated Model Building

Integrated Model

Institution Constraints

- Asset Classes
- Planning Horizon
- Threshold Constraints
- Cardinality Constraints
- Transaction Costs
- ...

Market Constraints

- Tax
- Regulatory Requirements
- Minimum Asset Reserve
- ...

Risk Constraints

- Downside
- VaR
- cVaR
- Variance
- Semi-variance
- Mean Absolute Deviation
- ...

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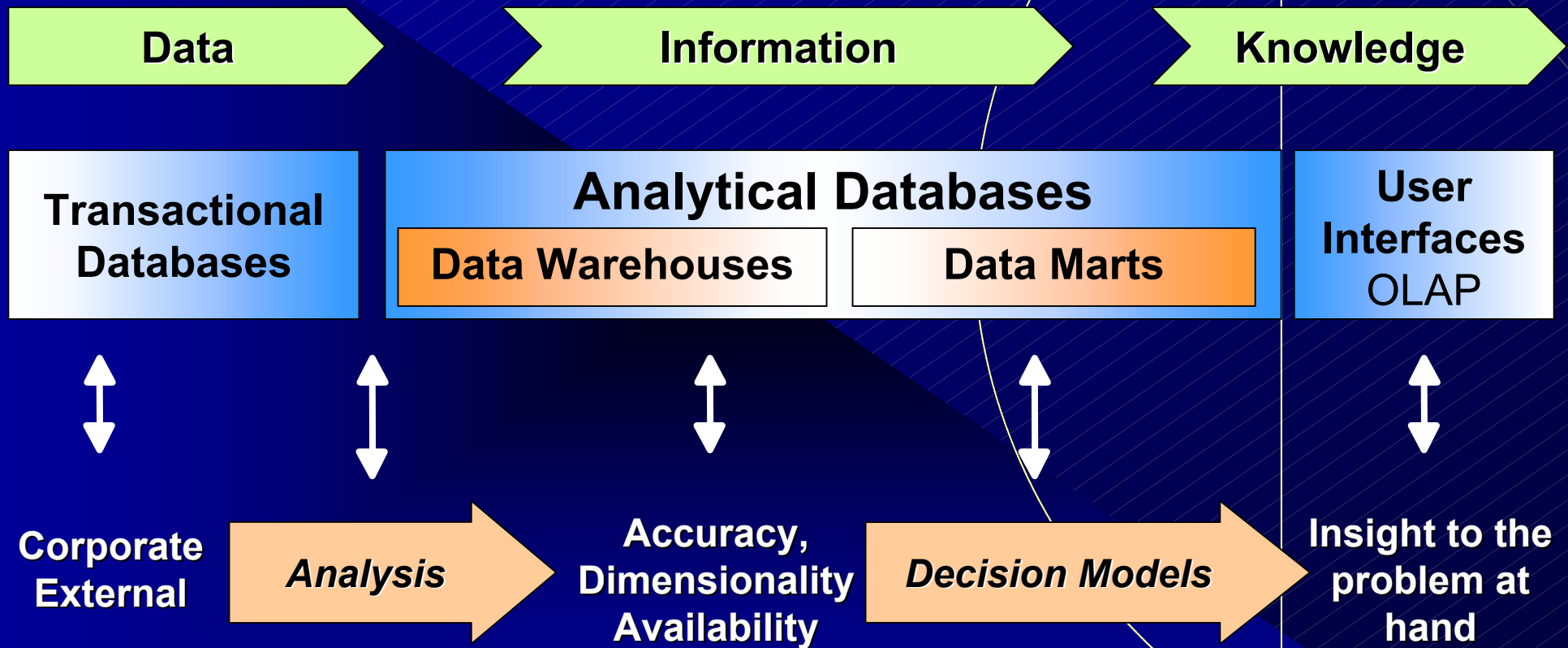
On-Line Analytical Processing

- ◆ With us since 1992
- ◆ Evolved into a User Interface feature for large data sets.
- ◆ Aimed for Analysts and Analytical models
- ◆ Highlights:
 - ◆ Multidimensional (therefore Realistic) Viewing of data
 - ◆ Hierarchical and Cross-Dimensional Data Operations (Roll-Up, Drill-Down, Slice, Dice)
- ◆ We use it for dynamic model instantiation
 - ◆ Feasibility of Models can be quickly investigated
 - ◆ Detailed results can be easily obtained
 - ◆ Sub-problems can be easily created

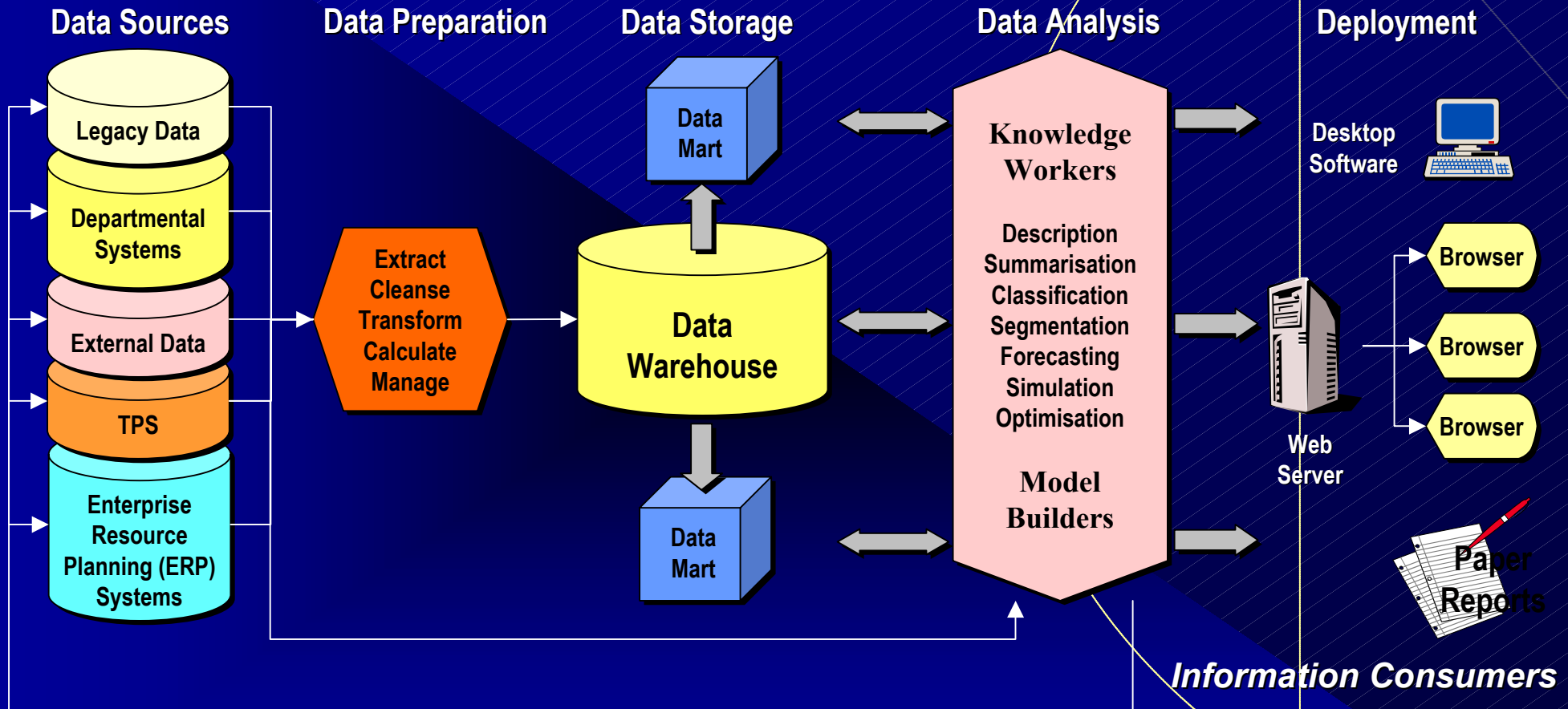
OLAP: how it fits in...

- ◆ OLAP evolved in parallel to Data marts and Data Warehouses
- ◆ Dimensional (data) modelling is common to all three:
 - ◆ Multidimensional user view of the world
 - ◆ Decisions and Business Analytics naturally view the world in terms of dimensions
- ◆ Mathematical Programming is naturally multidimensional (sets!)

Information Value Chain



The SPSS View : “Analytical Solutions Framework”



OLAP Operations in Decision Models

- ◆ Roll-Up, Drill Down, and Slice fundamental for model instantiation
- ◆ **Roll-up:** Aggregated models - feasibility studies, strategic decisions (Typical aggregation by Σ , μ . Others are possible)
- ◆ **Drill-down:** Detailed models - tactical decisions
- ◆ **Slice:** Sub-models - investigate weighting of model components

OLAP Operations in Decision Models

- ◆ **Model Investigation:**
- ◆ Summarise, Detail, or Filter data before model Instantiation
 - ◆ Requires careful data preparation and database design (e.g. alternative summary functions may be required for model consistency)
 - ◆ May require dynamic changes in the model
 - ◆ Industry standard connectivity with OLAP tools is still limited [XML may change that...]. We use Relational OLAP Schemas for most of our work, and SQL for OLAP operations.

OLAP Operations in Decision Models

- ◆ **Data Analysis:**
- ◆ Summarise, Detail, or Filter data following model instantiation
 - ◆ Most typical use of OLAP (e.g. Pivot Tables)
 - ◆ Requires basic preparation at the data level
 - ◆ Pretty straight-forward to do (unless there are complex aggregation and cross dimensional rules)
 - ◆ Not really useful and does not achieve OLAP's full potential in Decision Support Systems / Business Analytics

OLAP for DSS: Pros and Cons

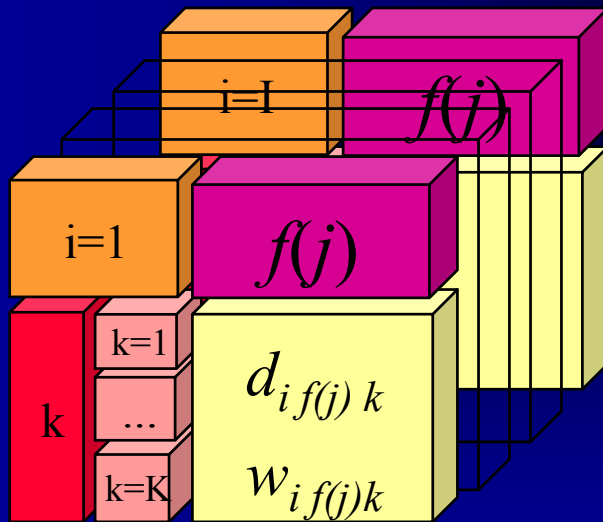
- ◆ Careful data and model preparation is necessary
- ◆ Results justify the preparation effort (in terms of using the DSS. Data values may not be as predicted/expected):
 - ◆ With few if any alterations the Data Mart drives the model investigation and/or data analysis
- ◆ High-level Dimensions can be used to store alternative data scenarios, and/or to drive alternative models

OLAP features in DSS Modelling

3 dimensions: i, j, k . 2 Vectors: d_{ijk}, w_{ijk} . Summary function: $f(j)$

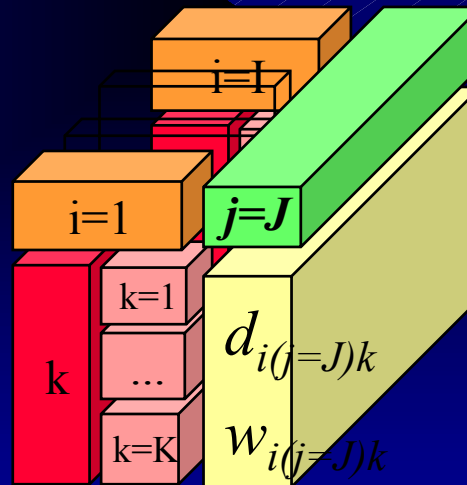
Summary Model

Roll Up over $j \equiv f(j)$



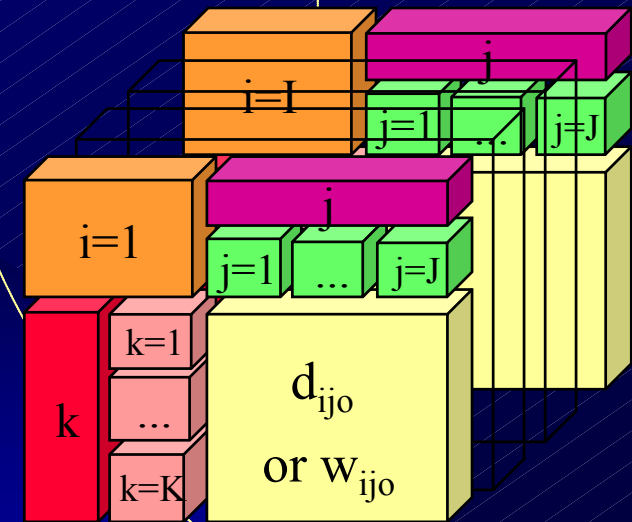
Sub-Model

Slice on $j, j=J$



Detail Model

Drill-Down on j



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Model Specification

- ◆ Planning Horizon: 12 months
- ◆ Initial Capital: \$1,000,000
- ◆ Additional Income per rebalance: \$10,000
- ◆ Rebalance Strategy: Twice at months 2 and 6
- ◆ Number of scenarios: 50
- ◆ Probability of occurrence: Uniform
- ◆ No liabilities (yet...)

Maximise

Terminal Wealth - Risk_t

Subject to:

Initial Holdings = 0

Wealth = investment * return

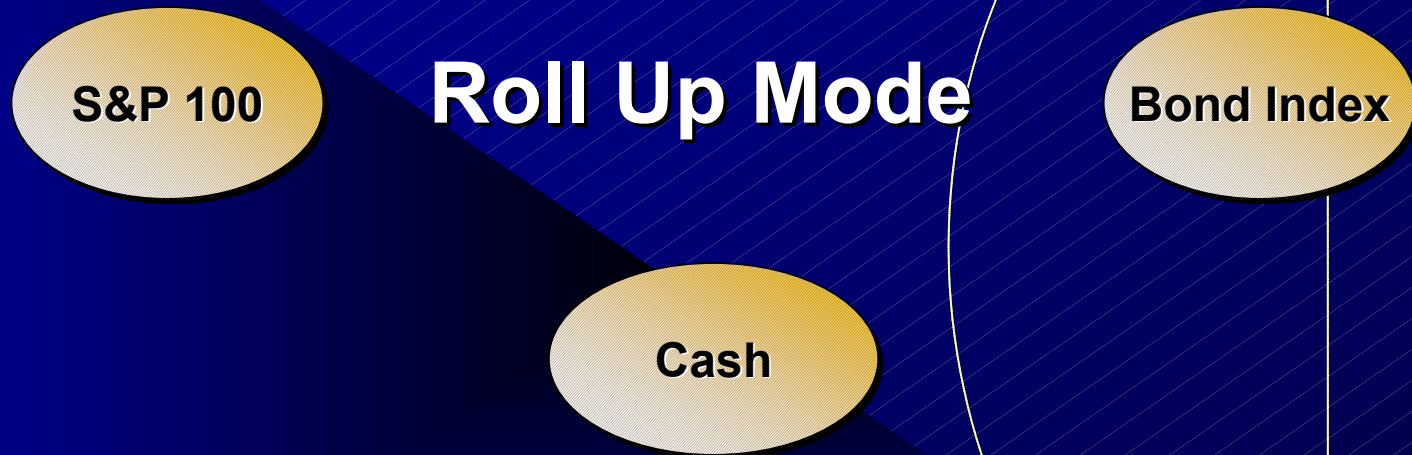
Fund balance constraint

Asset holding constraint

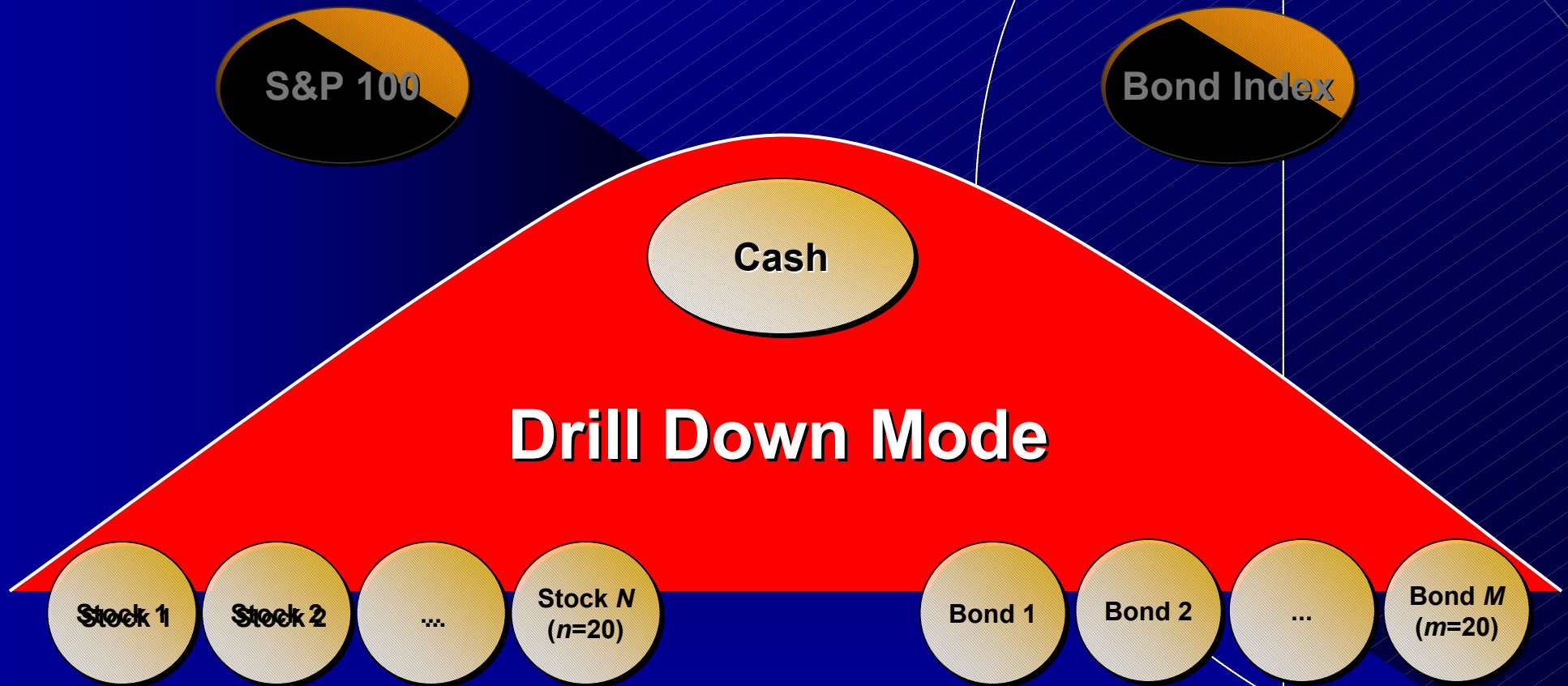
Transaction costs

Downside Risk (*Medium and Medium/High Risk Profiles*)

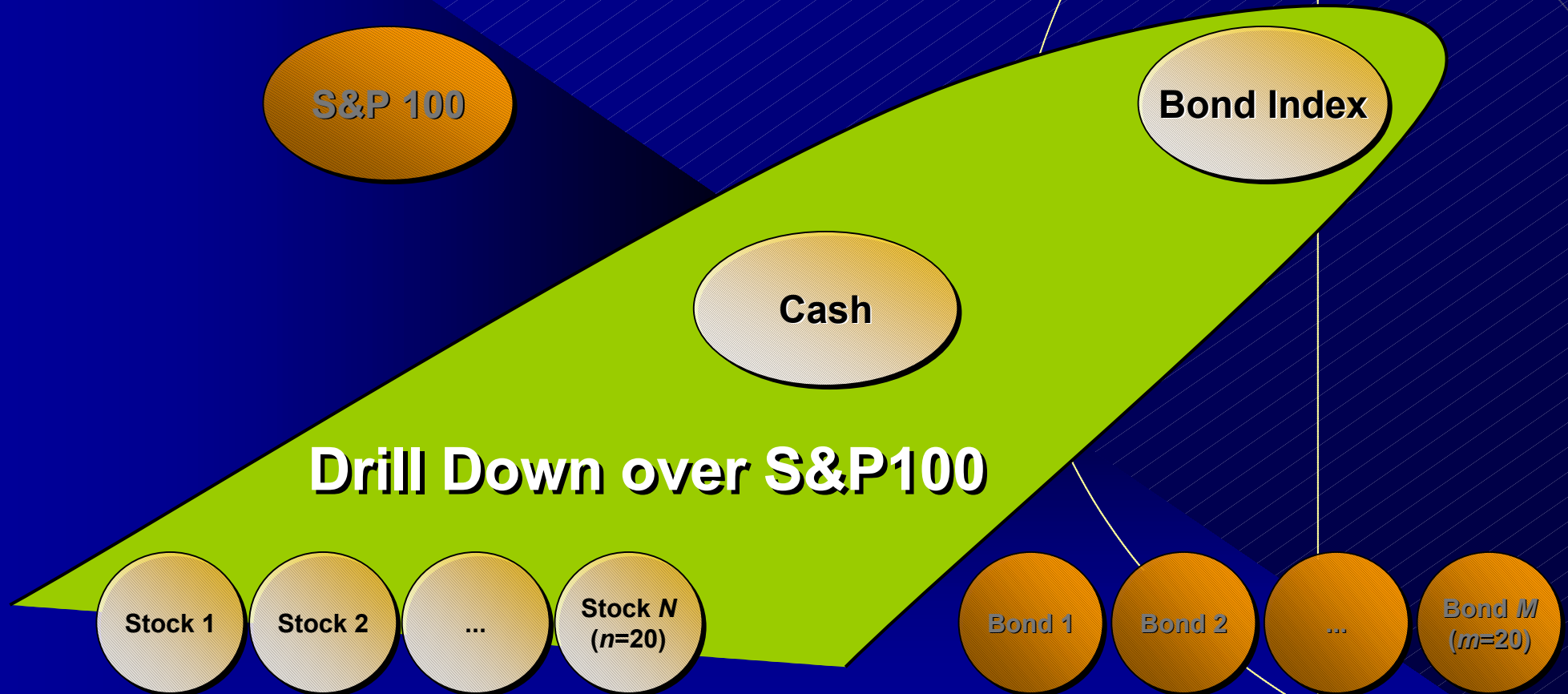
Model Specification – Asset Universe



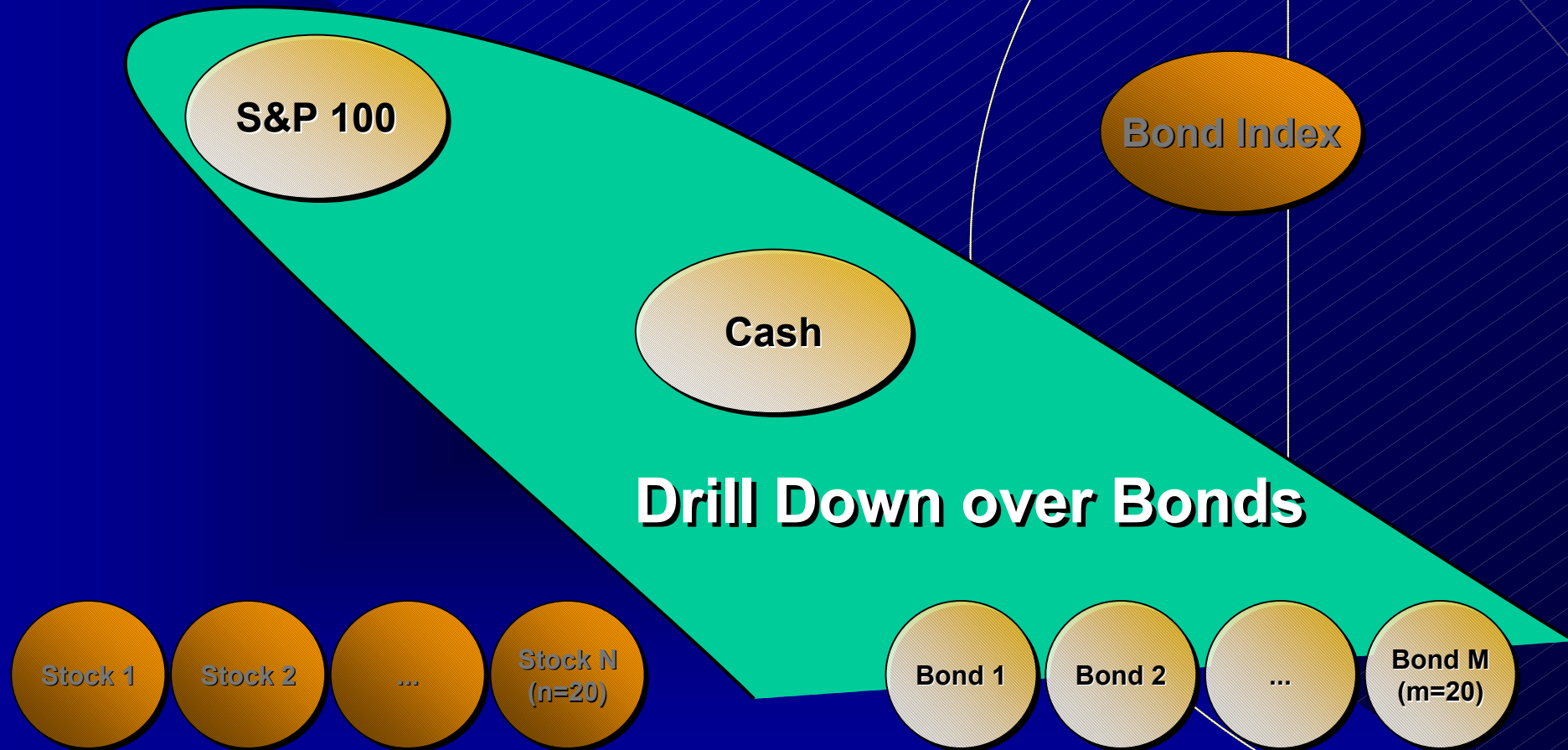
Model Specification – Asset Universe



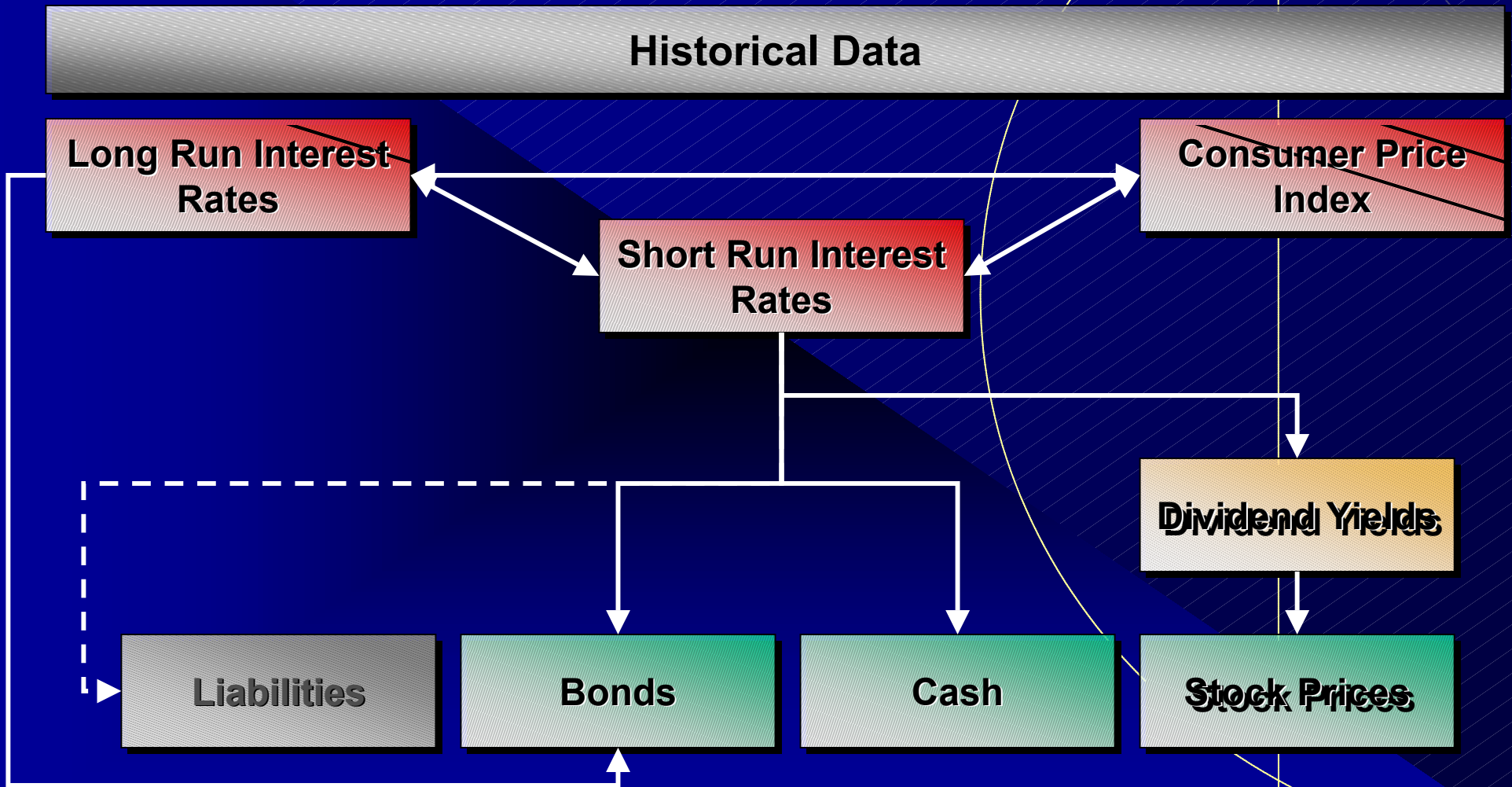
Model Specification – Asset Universe



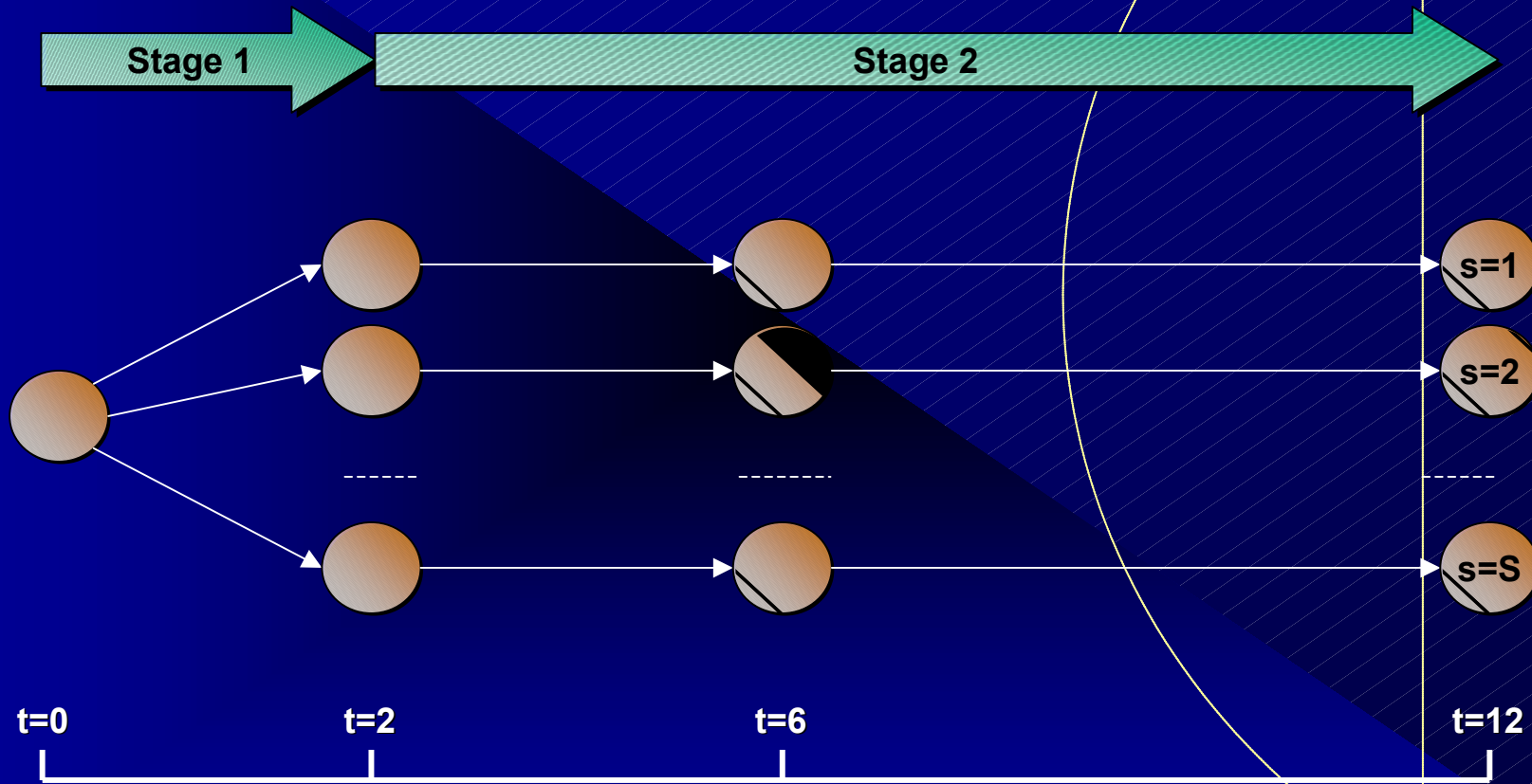
Model Specification – Asset Universe



An Integrated Scenario Generation System

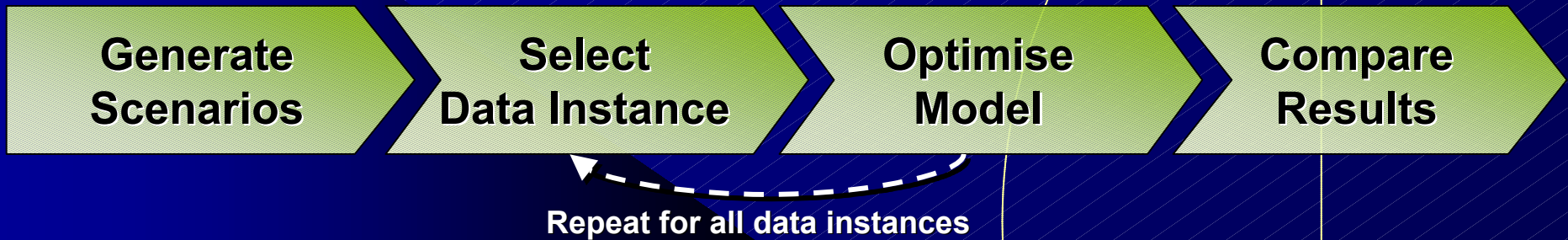


2-stage Stochastic Tree

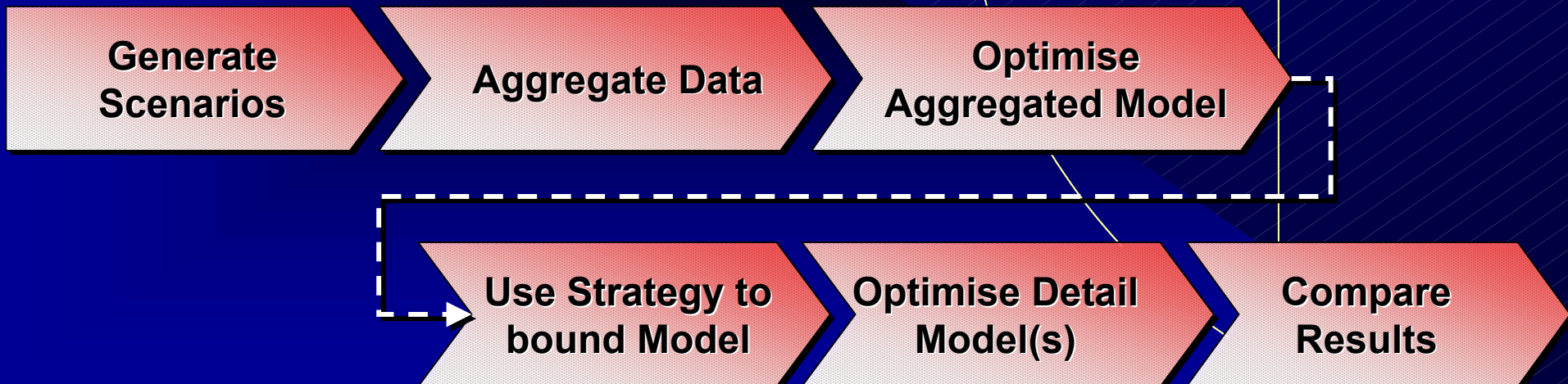


Our investigation

Currently



Where we want to go...



Formulation and Solving Strategy

- ◆ 1 Generic Model - 4 data instances:
 - ◆ Roll-Up Model: 3 Asset Classes
 - ◆ Semi-Drill-Down Models:
 - ◆ 20 Stocks + 1 Bond Index + Cash Investments
 - ◆ 1 Index (S&P100) + 5 Bonds + Cash Investments
 - ◆ Drill-Down Model:
 - ◆ 20 Stocks + 5 Bonds + Cash Investments

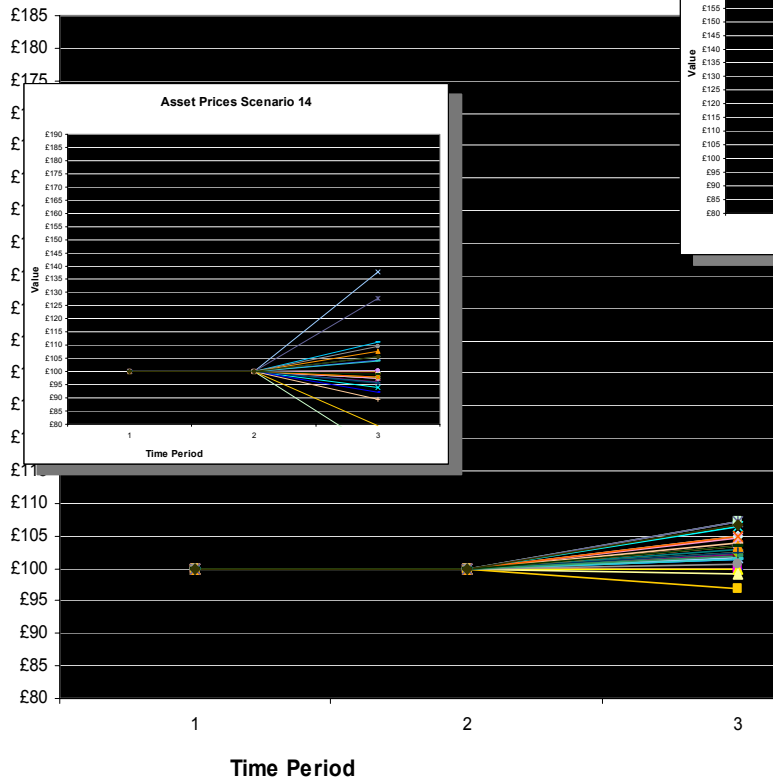
Work in progress, 3 challenges:
Scenario Generation
SP, Risk, and ALM Modelling
OLAP Integration

Currently:
Scenario Generation
Model Formulation

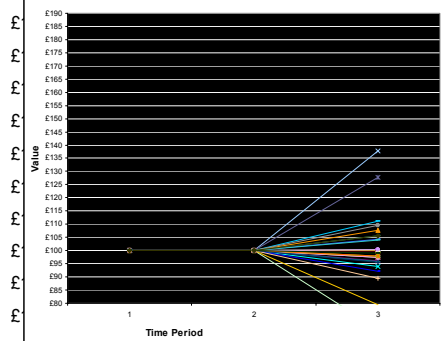
Upcoming:
OLAP Integration
...

Sample Scenarios

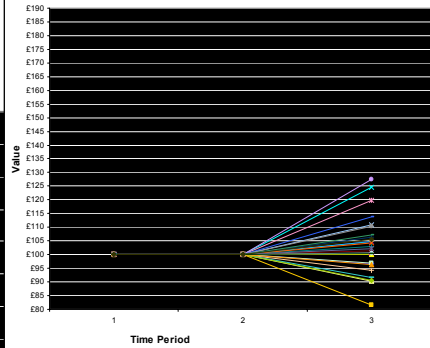
Asset Prices per Scenario (Average)



Asset Prices Scenario 14

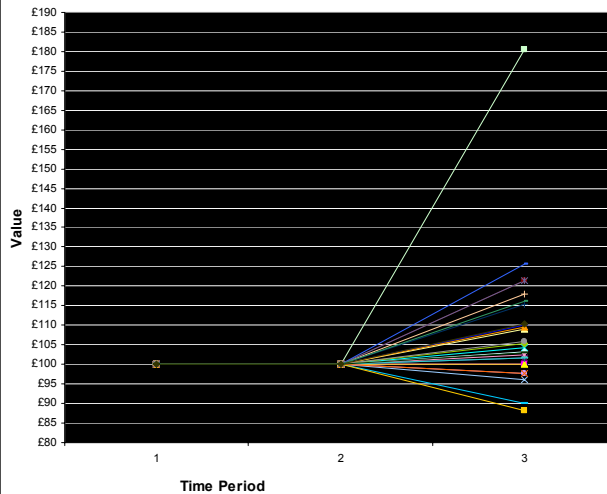


Asset Prices Scenario 27

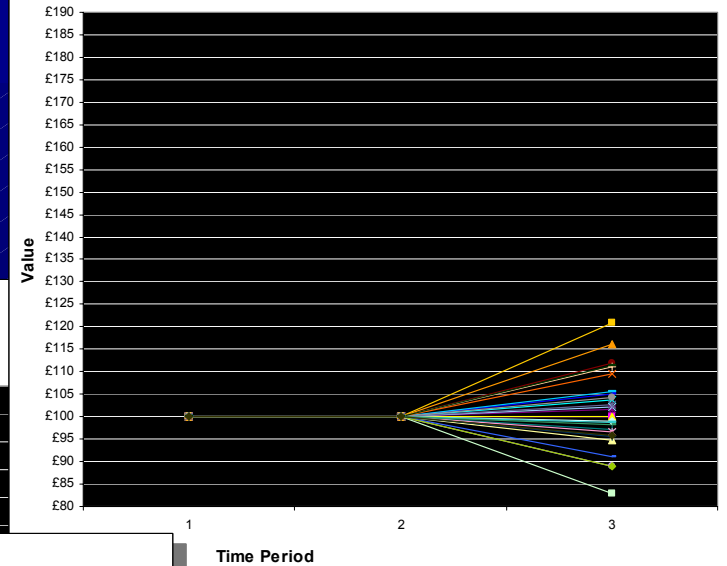


£170
£165

Asset Prices Scenario 17



Asset Prices Scenario 10



Time Period

2

3

Results so far...

- ◆ When Stocks are drilled-down, they constitute 100% of the portfolio.
- ◆ Stocks are the best investment for the risk-friendly investor.
- ◆ Results are comparable between Asset Classes and Drill-down over Bonds.
- ◆ Few bonds and cash for less-risky investors.
- ◆ Bonds and Cash are selected only when stocks are rolled up
- ◆ These results are specific to the scenarios generated (non-normally distributed)

Medium Risk Profile

Medium Risk	Roll Up	Roll Up	Roll Up
Asset Class	<i>today</i>	<i>month 2</i>	<i>month 6</i>
Stocks	44.00%	47.91%	46.06%
Cash	18.00%	22.14%	32.61%
Bonds	38.00%	29.95%	21.34%
Expected Wealth	£ 1,041,271.72	£ 1,095,642.21	£ 1,162,536.77
Medium Risk	Bonds DD	Bonds DD	Bonds DD
Asset Class	<i>today</i>	<i>month 2</i>	<i>month 6</i>
Stocks	36.00%	38.07%	40.22%
Cash	18.00%	24.26%	23.07%
Bonds	46.00%	37.66%	36.72%
Expected Wealth	£ 1,054,792.60	£ 1,118,297.23	£ 1,197,487.57

Medium/High Risk Profile

Medium/High Risk	Roll Up	Roll Up	Roll Up
Asset Class	<i>today</i>	<i>month 2</i>	<i>month 6</i>
Stocks	43.62%	44.16%	43.68%
Cash	21.68%	24.91%	30.19%
Bonds	34.70%	30.93%	26.13%
Expected Wealth	£ 1,033,833.99	£ 1,072,346.63	£ 1,118,990.73
Medium/High Risk	Bonds DD	Bonds DD	Bonds DD
Asset Class	<i>today</i>	<i>month 2</i>	<i>month 6</i>
Stocks	31.66%	31.63%	33.11%
Cash	24.62%	29.42%	23.04%
Bonds	43.72%	38.95%	43.85%
Expected Wealth	£ 1,036,978.72	£ 1,081,572.85	£ 1,131,287.28

Forthcoming Challenges

- ◆ Use OLAP techniques to fix the strategy on tactical decisions, with dynamic model readjustment (e.g. bounds may be necessary, etc.)
- ◆ Determine threshold risk profiles for investor
- ◆ Backtesting
- ◆ Fine-tune Scenario Generation