

# Why Do Models Have Limitations?



13 November 2015

# Model Limitations – Why do we Care?

#### A great deal of focus on model limitations in Solvency II

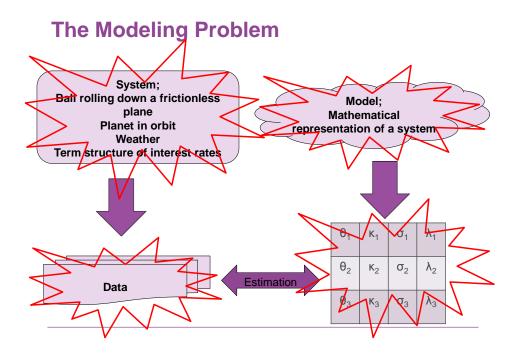
#### Why does the regulator care?

- Concern that market outcomes will not be adequately captured leading to insolvency
- · A desire that risks are adequately priced into businesses
- · A perception that models contributed to the last/current crisis/crises
- Model risk

However all models have limitations – everyone always knew this The question that needs to be addressed is what are the *material* limitations?

- · The answer is likely to differ from user to user
- · In most cases quantifying the model risk is only partially possible

This talk will look at why models have limitations and ask does it matter?





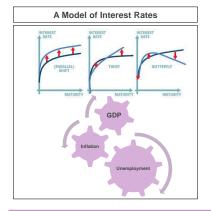
# **The System**

Reality is Reality and Models are Models



#### The Extent of Limitations Depend on the System

- Most systems are highly complex
- In building models we substitute this complexity for something tractable
- · Most financial models are a representation of effect rather than cause
  - · Even "fundamentals" are not really fundamental





# The Limitations on the Model Depend on the State of System

- · Models are best suited to modeling markets which are "free" and liquid
- Models cannot be expected to perform as well and may fail when "structural" change occurs
- · Models cannot easily capture a range of "artificial" effects
  - Quantitative easing
  - Geo political effects (e.g. Break up of the Eurozone)
  - Economic restructuring
- · A failure of a model does not (automatically) make it misspecified



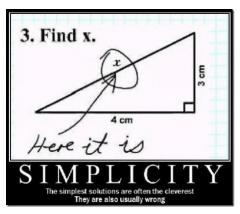
# **Model Specification**

The Search for Parsimony

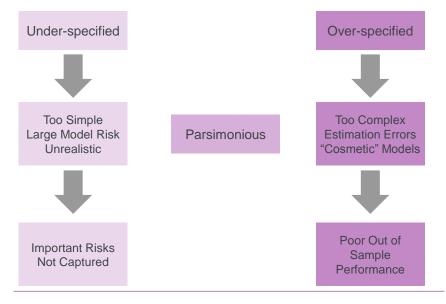


# **Ockham's Razor**

- "Simple models are better models"
- This is actually not an accepted definition
  - Entities must not be multiplied beyond necessity
  - We consider it a good principle to explain the phenomena by the simplest hypothesis possible (Ptolemy b. AD90)
  - We are to admit no more causes of natural things than such as are both true and sufficient to explain their appearances (b. I. Newton 1642)
- What Ockham's Razor is really talking about is parsimony
  - Smallest number of factors to explain the maximum amount of variance



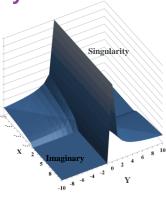
Source: Google Images



#### What is Parsimony and why is it important?

# The Limitations of Parsimony

- Parsimony reduces the complexity of the system with the minimum loss of information
- Models must be mathematically tractable parsimonious Restricting ourselves to the tractable parsimonious models however engenders limitations
  - They tend to produce smooth continuous distributions
  - The model may contain boundary conditions and singularities
  - · We may want the model to do something which is outside of the parameter space
- Why not just add more factors then?
  - · We may solve one problem for others to appear
  - · A model that can do everything probably will
  - The additional factors cannot be estimated they are just noise (False Precision)



Source: Conning RCMS



# **Data Limitations**

$$\frac{2^{HS}}{2^{HS}} = \frac{1}{2^{HS}} =$$

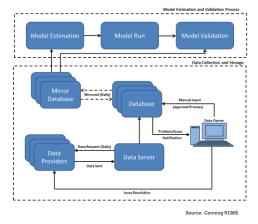
## **Data Issues**

- Accuracy
  - · How noisy is the data
  - · Accuracy of the data is often difficult to assess
  - · Using multiple sources does not solve the issue
  - Data corruption
- Completeness
  - · Often time series data is too short for valuing long term risks robustly
  - Data granularity
- Appropriateness
  - Expost vs. Exante
  - · End of day data biases
  - · Selection bias particularly within index data is also a key consideration

# **Tackling Limitations in Data**

Data limitations can be tackled on several fronts

- Accuracy
  - Using long histories of data can limit the effect of a small number of spurious points
  - Using noise reduction techniques to estimate the model from the data
  - Reduce manual processes
- Completeness
  - Consider augmenting/splicing multiple data sets
  - · Extrapolation and interpolation
- Appropriateness
  - · Ensure that data used is specific to the asset class/local being modeled
  - · Have a consistent approach for when data is not available
  - Expert judgment



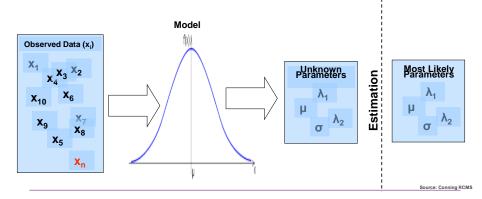


# **Model Estimation**



### **Estimation**

- · Even with good data how the model is estimated may introduce limitations
- Often the most useful models do not have parameters and factors which are directly observable (short rate models, stochastic volatility, jumps)
  - · What's more the models are often "continuous time"
- · Analytical techniques must be used to link the model world to the real world



#### Kalman Filter and MLE – Robustness of estimates

- We can quantify how good these methods are
- · Method:
  - Fix the model parameters to some known values
  - Simulate yield curves for 10 years at monthly frequency
  - Take 250 simulations and run KF + MLE on each one to recover estimates of the model parameters
  - Compare the parameter vector distributions to the input parameters
- The results are good although they will be biased to an extent by:
  - The optimizer used
  - Discretization error

 $dy_1(t) = \kappa_i(\theta_1 - y_1(t))dt + \sigma_1 \sqrt{y_1(t)}dW_1(t),$ 

 $dy_n(t) = \kappa_n(\theta_n - y_n(t))dt + \sigma_n\sqrt{y_n(t)}dW_n(t),$ 

Parameters	CIR					
	Actual Mean values estimat		Standard deviation			
	CIR					
Parameters	Actua values		Mean estimate		Standard deviation	
ĸ	0.10	0.1	0.141		0.053	
θ	0.05	0.0	0.041		0.013	
σ	0.075	0.0	0.075		0.005	
λ	-0.40	-0.4	-0.437		0.042	
$\lambda_1$	-0.15	-0.193	0.0	078		
$\lambda_2$	-0.10	-0.125	-0.125 0.0			
$\lambda_3$	-0.05	-0.074	0.0	0.051		

Source: J. Bolder, Affine Term-Structure Models Theory and Implementation



# Model Usage

**Behavioral Aspects** 

$$\frac{2^{HS^{e}}}{2^{HS^{e}}} \frac{1}{2^{HS^{e}}} \frac{1}{2^{HS^{$$

# **Choice of Metric**

- In many cases statistical models are used to produce a single or limited number of metrics to describe risk
- · Which metric is chosen will carry its own limitations
  - Volatility
  - VaR

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- cVaR, TVaR, Expected Shortfall
- As does the quantile
  - 99%, 99.5%
- Depending on the distribution this may give quite different views of risk
- In Insurance we exist within a regulatory framework with a "single metric" definition of risk

# Paradigms – Why do we repeatedly mis-apply models?

- In many areas paradigms develop which become entrenched
- The physicist Thomas Kuhn suggested how such paradigms develop
- Inauguration
  - An event occurs or a new concept is introduced which gives rise to a paradigm shift
- Vigor
  - The paradigm shift gives rise to whole new areas of research, new disciplines and practical applications
- Dominance
  - The paradigm comes to dominate activity in the area in which it occurred
- Revolution/Evolution
  - An event occurs which shows the limitations of the paradigm and new ideas develop to replace it





Case Study

Multifactor Interest Rate Models

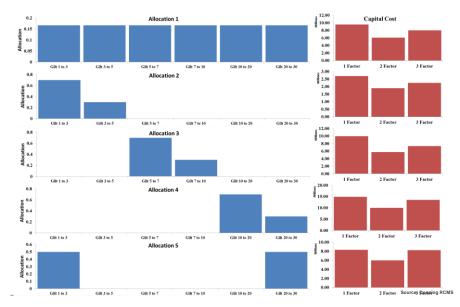


# **Do Model Limitations Matter?**

- Test using a 3 Factor, 2 Factor and 1 Factor Model of Interest Rates
- Model used is a multi factor Cox, Ingersoll, Ross Model
- A "through the cycle" parameterisation is used
  - · Estimated from 55 years of data
  - · Starting point is the year end 2014 Gilt curve
- · The 3 Factor model is estimated first
  - The 2 and the 1 factor model are then estimated using the same data
  - An additional constraint is put on the mean and volatility of the medium horizon yields (5 Years) and returns 3F=2F=1F
- · What impact does the number of factors have on capital cost?



## **Test Allocations**



#### **Results**

## Summary

- There are many reasons that models have limitations some of which have been identified
- · Understanding limitations are an important element of solvency II
- · Often we are inclined to "solve" limitations
  - Doing so may engender new limitations
- · There are other ways to assess the impact of limitations though
  - · What if analysis
  - · Stress testing
  - Discussion

It is the process of developing an understanding of model limitations which adds the most value to a risk management process, and identifies key risks and opens dialogue on how to mitigate those risks.



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