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# **Economic Scenario Generators**

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# Agenda



- What is an ESG
- Types of ESG
- Why do we need an ESG
- Risk neutral ESGs
  - Models
  - Calibration
  - Validation
- Challenges in Indian markets

# Introduction to Economic Scenario Generator



- Future is unknown
- We may have expectations about the future but we are never certain about it
- An ESG is a tool which
  - Uses Monte Carlo simulation to
  - Generate numerous simulations of economic variables
  - Over multiple time periods
- Average of the simulations converge to our expectation

# Introduction to Economic Scenario Generator



# Types of ESG

## Risk Neutral (RN)

- **Market consistent**: Parameters of underlying models are calibrated such that *economic scenarios* are *consistent with market prices*
- **Risk neutral**: Scenarios are modeled ensuring that no arbitrage allowed. All financial instruments will have the same *expected return* which is *equal to the risk free rate*
- Individual scenarios results do not hold any significance
- Used for pricing and valuation only
- Not intended to reflect real world expectations

## Real World (RW)

- **Subjective**: Economic scenarios modeled to reflect subjective views about the future evolution of the markets
- **Not market consistent**: Economic scenarios are not consistent with current market prices
- Incorporate risk premia in asset returns
- Individual scenarios can be used for analysis
- Used for activities which require realistic forward looking projections

# Why do we need an ESG?



Market consistent valuation of options & guarantees

RN

Formulation of ALM/ investment strategies

RW

Risk management/ economic capital calculation

RW

Business planning/ Capital planning

RW

Pricing

RN

# Why do we need an ESG?



As per APS 10 , Embedded Value should

- Allow for time value of Financial Options & Guarantees
- Allowance should be based on stochastic techniques
- Economic assumptions should be in line with capital market prices of similar traded cash flows  $\longleftrightarrow$  Market consistent
- As identical traded options may not exist, we need a Market Consistent/ Risk Neutral ESG

# Types of options & guarantees embedded in life insurance products



Non-linear payoffs/ guarantees need to be valued using an ESG

Examples

- **Minimum return guarantee** in participating/Unit linked products:
  - Guarantees in par products are non-linear
  - Upside shared between SH and PH
  - Downside fully borne by SH
- **Surrender option**
  - Similar to an American option
  - Can be exercised at any point of time during the contract depending on the perceived value of the option



# Types of options & guarantees embedded in life insurance products



## Examples

- **Paid-Up option**
  - Similar to Bermudan options
  - Can be exercised at premium payment dates
  
- Valuation of options is tricky as it requires assumptions about “Option exercise strategy/ policy holder behavior”

# Risk neutral ESGs

Selection of asset models

Calibration of model parameters

Generation of economic scenarios using  
Monte Carlo simulation techniques

Validation

# Asset Models



Very generically, all asset models are of the form:

- $dS = a(t, S_t) * dt + \sigma(t, S_t) * dW_t$

Where  $a$  &  $\sigma$  are the drift and diffusion functions and

$W_t$  is a Weiner process

- $W_t$  has Gaussian increments, i.e. the distribution of  $W_t - W_s \sim N(0, t-s)$
- The increments are independent of each other
- $W_0 = 0$

# Asset Models: Interest rate models



## Interest rate models

**SHORT RATE MODELS**: Model the behavior of instantaneous spot/ forward rates

**MARKET MODELS**: Model the behavior of forward rates observed in the market

## Short rate models

### **ONE FACTOR:**

Example – Hull White 1-F model

$$dr_t = [\theta(t) - ar_t]dt + \sigma*dW_t,$$

where  $a$  and  $\sigma$  are positive constants and  $\theta(t)$  is chosen so that the model exactly matches the term structure of interest rates

### **TWO FACTOR/ MULTI FACTOR**

# Asset Models: Interest rate models



## Short rate models

### TWO FACTOR:

Example – Hull White 2F model/ G2++ model

$$dr_t = (\theta(t) + u(t) - ar_t)dt + \sigma_1 * dW_{1,t}$$

$$du(t) = b * u_t * dt + \sigma_2 * dW_{2,t}$$

$$dW_{1,t} * dW_{2,t} = \rho * dt$$

where  $a$ ,  $b$ ,  $\sigma_1$ ,  $\sigma_2$  and  $\rho$  are positive constants and  $\theta(t)$  is chosen so that the model exactly matches the term structure of interest rates

1-F versus 2-F models

- Easier analytical tractability in 1-F models
- However, the resultant spot rates for all maturities are perfectly correlated with each other. Thus a **one factor model allows only for parallel shifts of the yield curve** and no shape changes are possible.

# Asset Models: Interest rate models



## Market Models: Libor Market Model

### Libor Market Model:

Most widely used markets in developed markets

LMM models forward rates which are observable in the market

Each forward rate  $F(t, T)$  follows a process where the drift is dependent on the other forward rates

Correlation between different forward rates is also allowed for.

Leads to a better fitting of volatility structure of interest rates

Market models versus Short rate models

- Market models are relatively difficult to implement
- Market models need a lot many data points for calibration
- Allow for a better fitting of volatility structure of interest rates

# Asset Models: Equity



## Black Scholes Merton model

$$ds_t = \mu * S_t * dt + \sigma * S_t * dW_t$$
$$S_{t+1} = S_t * \exp[(\mu - \sigma^2/2) * t + \sigma * W_t]$$

Lognormal is the simplest model for Equity prices.

It assumes a constant volatility structure

Unable to replicate market prices of out of the money options

*There is a trade off between the complexity of model & the goodness of fit.  
Models need to be chosen based on the requirements of the task in hand*

# Risk neutral ESG Calibration



Calibration is the process by which the parameters of the chosen models are estimated.

Objective Calibration criteria: Model fits the observed market prices of options

Options used: equity calls/puts, interest rate caps/ floors, swaptions

Model parameters are usually calibrated using

- Analytical expressions for option prices (for simplistic models)
- Numerical methods – Building Trinomial trees (for most models)
  
- Illustration for building trinomial trees has been given in the paper and is also available online
  
- Codes\ Packages for calibration exist in open source softwares like Python & R



# Generation of simulations & Validation



## Generation of simulations:

- Monte carlo simulation techniques applied on the calibrated models
- For simulating joint behavior of economic variables
  - Correlation between asset classes is estimated based on historical data
  - Cholesky decomposition is used to generate correlated random numbers

## Validation:

1. Risk neutrality – Martingale test: Average of discounted value of any asset over the simulated paths should be equal to current market price of the asset

$$\sum_{i=1}^{N=\text{Simulated ESG paths}} \frac{D_{i,t} * (1 + R_{t,i}^{cum})}{N} = 1$$

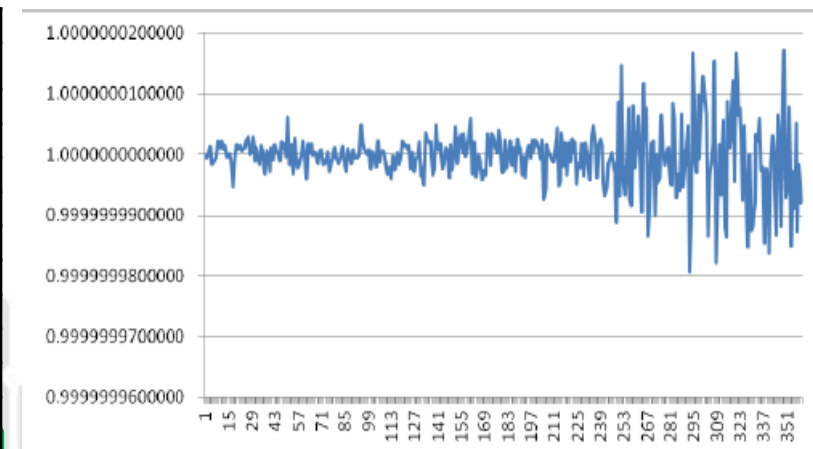
# Validation



Cumulative Returns on NIFTY						
SCENARIO\ TIME	0	1	2	3	4	5
1		103%	122%	142%	115%	81%
2		89%	88%	125%	134%	146%
3		132%	121%	153%	166%	193%
4		110%	113%	131%	135%	165%
5		112%	118%	104%	115%	112%
6		100%	104%	108%	128%	134%
7		106%	145%	217%	235%	197%
8		105%	111%	99%	119%	187%
9		125%	148%	161%	185%	206%
10		98%	100%	106%	89%	100%

Path wise Discount Factors						
SCENARIO\ TIME	0	1	2	3	4	5
1	1	95%	87%	84%	81%	79%
2	1	91%	80%	70%	62%	54%
3	1	89%	79%	74%	71%	70%
4	1	92%	88%	78%	72%	66%
5	1	92%	85%	76%	73%	72%
6	1	91%	75%	65%	59%	55%
7	1	90%	78%	68%	61%	55%
8	1	92%	86%	77%	72%	67%
9	1	94%	91%	82%	69%	58%
10	1	94%	86%	78%	72%	65%

Martingale Test Statistic						
SCENARIO\ TIME	0	1	2	3	4	5
1		98%	107%	119%	93%	64%
2		81%	70%	88%	83%	80%
3		118%	96%	113%	118%	135%
4		101%	99%	102%	97%	109%
5		103%	101%	79%	84%	81%
6		92%	78%	70%	75%	74%
7		96%	114%	149%	143%	108%
8		97%	96%	77%	85%	126%
9		118%	135%	132%	127%	120%
10		92%	86%	83%	64%	65%
Average*		1.00	1.00	1.00	1.00	1.00



# Validation

## Market consistency/Goodness of fit test:

- Comparison of prices of traded instruments
  - computed using ESG simulation output
  - actual traded prices

$$P(t = 0) = \frac{\sum_{i=1}^{N=\text{all paths in ESG output}} D_t X_t}{N}$$

- LHS is the actual price of an option
- RHS is the price computed using ESG output (Average of the discounted value of option payoff)

# Challenges in Indian markets



- Data required for risk neutral ESG calibration
  - Yield curve
  - Equity Implied option volatility – NIFTY options
  - Implied volatility on interest rate options – Swaptions, Interest Rate caps & floors, bond options
- Challenges
  - Equity implied option volatility:
    - Only short tenure options are available
    - Implied volatilities of options varies by tenure of the option
  - Interest rate implied option volatilities:
    - Interest rate options are traded only OTC
    - Data is thin and difficult to obtain

# Challenges in Indian markets



- Possible solutions - Equity
  - **Assume a constant volatility**: leads to an over-estimation of short dated options and under-estimation of long date options
  - **Functional form for implied volatility**:
    - Use observed implied volatility data
    - Use a long term volatility assumption (Based on realized long term volatility)
    - Impose a functional form for the volatility term structure
    - Interpolate/ Extrapolate volatility for tenures to be used for calibration

# Challenges in Indian markets



- Possible solutions – Interest Rate volatilities
  - Bond options are already trading on exchanges – short term only
  - Use of data from other developed markets:
    - Use the implied volatility surface from developed markets
    - Compute the relative value factors of implied versus realized volatility
    - Apply these factors to realized volatility of forward/ swap rates
  - Directly use the forward/ swap volatilities observed in Indian markets and fit volatility estimation models like GARCH to estimate forward looking volatilities