## **Session 6: Financial market returns**

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### Goals

The importance of returns
The distribution of returns
Value at Risk (VaR)

## The importance of returns

### **Prices and returns**

- All of us know the time-series of prices  $p_t$ .
- Finance requires a major shift in focus away from prices to the time-series of *returns*. For right now,

$$r_t = 100 \left(\frac{p_t}{p_{t-1}} - 1\right)$$

# Why returns is the right way to think about finance

- Prices are not comparable across products.
- Prices can move dramatically for non-informative events: e.g. stock split.
- Most asset prices trend up, and it is not meaningful to talk about the distribution of something that is not stationary.

## prices



# **Comparability comes by shifting to returns**



## **The concept of returns is general!**

Returns of equities, equity indexes, currencies, commodities: all are obvious.

Interest rates: Shift from thinking about interest rates to *returns* on bonds.

## The distribution of returns

## What do we know about prices

- Prices can move up and down,
- **But they can't become negative.**
- There can be very large positive moves.
- **But the returns can't be worse than -100%.**

## **The right parametrisation**

$$r_t = 100 \log_e(p_t/p_{t-1})$$

- If  $p_t \to 0$ , then  $r_t \to -\infty$ .
- If there is a massive upswing in prices,  $r_t \to +\infty$ .
- Fact:  $p_t/p_{t-1}$  is roughly lognormal;  $r_t$  is roughly normal.

### **Raw versus continuously compounded returns**

Fact: As  $x \to 0$ ,  $\log_e(1+x) \approx x$ , so for small values of returns,  $\log_e(1+r) \approx r$ . Example:  $\log_e(1.03) = .0295588.$ 

This can all be confusing!! 3% or 1.03 or 0.02955 or 2.955. Choose one way of measurement (2.955) and rigidly stick to it.

For weekly or monthly returns, values of *r* are bigger, so whether you log or not makes more of a difference. For small time horizons it matters less.

## Looking at real-world returns

## they? returns : now 'normal' are



## summary statistics

	Nifty	S&P 500	INR/USD
$\mu_{\rm price}$	973.4384	742.0828	39.4443
$\sigma_{\rm price}$	300.4205	363.6260	5.9551
$\mu$ returns	0.0705	0.0622	0.0216
<i>σ</i> returns	1.9984	1.0008	0.2817
1 <sup>st</sup> Quartile	-0.9762	-0.4214	-0.0280
Median	0.0507	0.0622	0.0216
3 <sup>rd</sup> Quartile	1.1000	0.5719	0.0560

### An across-markets view

#### 2003-04 vol

Nifty	1.43
COSPI	1.47
1-year ZC bond	0.32
10-year ZC bond	0.47
GOI bond index	0.27
Gold	0.92
Silver	1.42
INR/USD	0.19
INR/EUR	0.77

## Value at Risk (VaR)

## **Defining VaR**

Suppose  $r_t \sim f(r)$ .

VaR v is a r<sub>t</sub> defined in the following manner: VaR is the value v at a p level of significance using the equation

$$\int_{-\infty}^{v} f(r)dr = 1 - p$$

# A 90% VaR on a standard normal distribution



### A 90% vak on a normal distribution with Nifty parameters



## **Two CDFs and the 10% cutoff**



## References

- http://www.economagic.com has some useful data.
- http://www.mayin.org/ajayshah
- http://www.igidr.ac.in/~susant
- PHILIPPE JORION. Value at Risk : The Benchmark for Controlling Market Risk. McGraw Hill, 2000, 2nd edition