

Rolling Hills

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The problem (i)

- Can't tell you yet
- Stylised problem discussed here
- Standard approach used for consulting but this amusing viewpoint was tangential

The problem (stylised)

- How much capital for extreme stock price movements?
- Can we get a good model of extreme returns?
 - Extreme returns rare - need lots of data
 - Distributions change over time
 - Fat tails

Stage one

- Reduce fat by conditional variance
- Model variance by GARCH
 - log return at time $t := r_t$,
 - standardised return $:= s_t$,
 - volatility estimate at time $t-1 := \sigma_{t-1}$.
- $s_t = r_t / \sigma_{t-1}$
- $\sigma_t^2 = k\sigma_m^2 + j\sigma_{t-1}^2 + (1-j-k)s_t^2$
- σ_m^2 is a long term central value to which variance of return reverts at a speed related to k .
- We are assuming the average return is zero over small time intervals to avoid estimating the nuisance parameter of the mean return

Stage one (ii)

- Standardised returns are still fat-tailed
- Skew is close to zero
- Overall std dev is close to 1.0
- Model the tails

EVT – a quick intro

- Tails of distribution of returns are Generalised Pareto Distribution (GPD)
- $F(x) = 1 - (1 + kx/s)^{-1/k}$
- x is sufficiently large and positive, it is presumed to be larger than some threshold u

EVT

- $s > 0$ and is a scale parameter – the larger the value of s , then the more spread out are the values of x
- k is a shape parameter – as k decreases ($1/k$ increases), the tail of the distribution becomes fatter
- $1+kx/s > 0$
- **No standard notation available**

Shape parameter

- Hill estimate is used for the shape parameter:
- $k = \Sigma \ln(x/u) / n$
- where:
- n is the number of observations where x is more extreme than u ,
 - summation is only over those extreme values

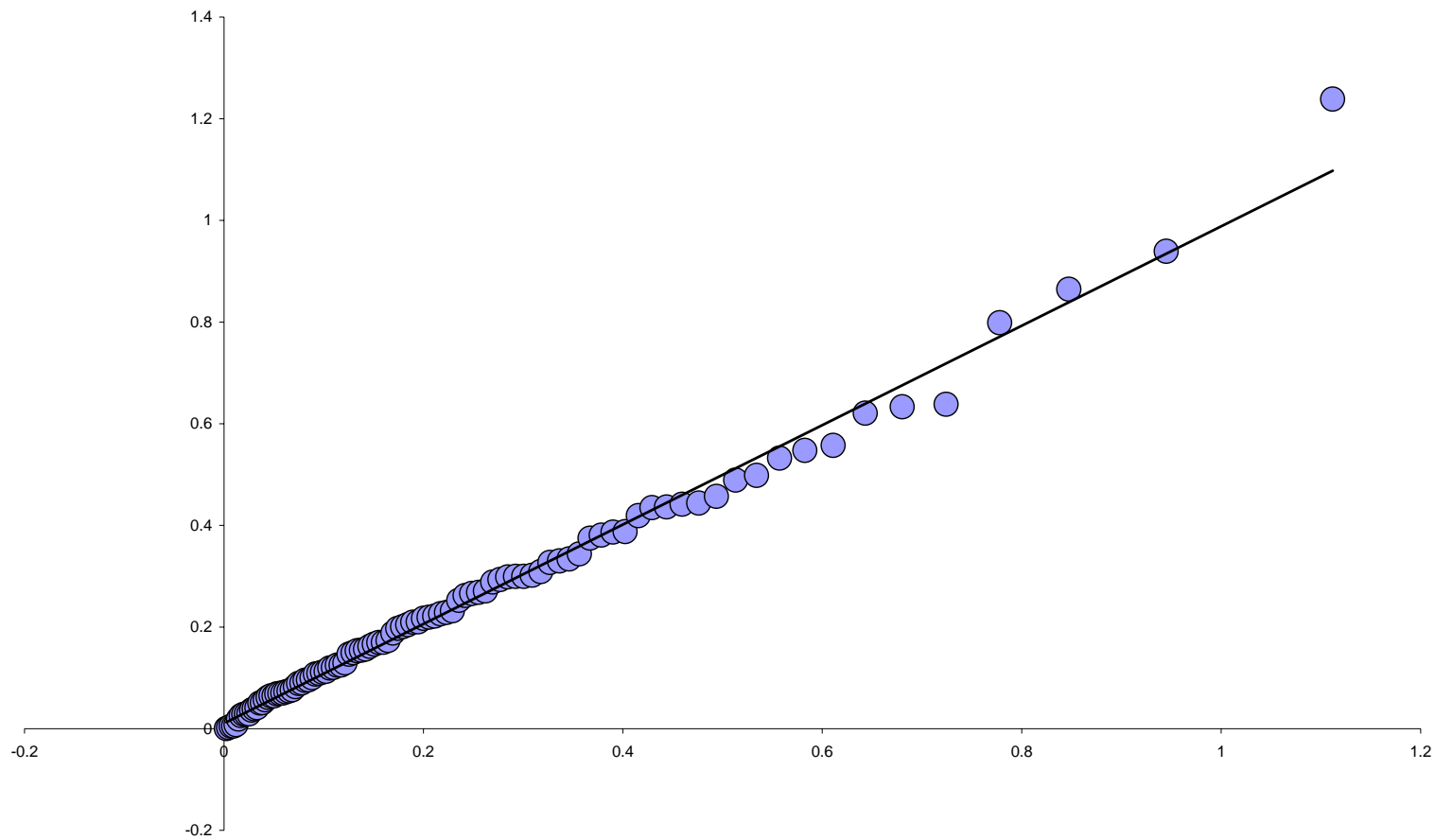
Scale parameter

- $s = k(\Sigma(x-u)^k / n)^{1/k}$
- The shape parameter governs how fast the tail decays – like $x^{-1/k}$
- The scale parameter gives us a measure of the typical size of the exceedances
- Both are important in the determination of the VaR

Quantile Plot (Q Plot)

- If GPD is good, log relative rank versus log exceedance should be a straight line.
- Q Plot drawn to enable judgement
- We may only have a few observations in the tail, and so not much statistical power – eyeballing may be the best test.

BHP QQ plot



Value at Risk

- Let p be the probability that an observation is in the tail.
 - probability of value more extreme than u is p
- the percentile at probability level q is
- $VaR_q = u + T_q$
- $T_q = (s/k) [(q/p)^{-k} - 1]$
- For good estimates need $q \ll p$.

Typical uses

- Pick a tail probability like 5% or 2.5%
- Estimate 1% Var
- Estimate expected shortfalls

Problems

- How do parameters change?
- Every new data point:
 - Re-estimate percentiles
- Give recent data extra weight?
 - Difficult to calculate percentiles

Exponential weighting

- Typically used as a simple GARCH process
- $\sigma_t^2 = k\sigma_m^2 + j\sigma_{t-1}^2 + (1-j-k)s_t^2$

EW-EVT

- Model tail standardised log returns
- Don't choose tail by p eg $u=F^{-1}(2.5\%)$
 - Choose tail by cutoff point eg $u=-2$
- Tail probability is now variable
- Choose long decay factor
 - $0.996 \rightarrow$ avg age = 250 days

Tail probability

- $p_t = jp_{t-1} + (1-j)I_t$
- I_t is:
 - 1 if we're in the tail
 - 0 otherwise
- probability goes up if we're in tail, goes down otherwise

Shape parameter

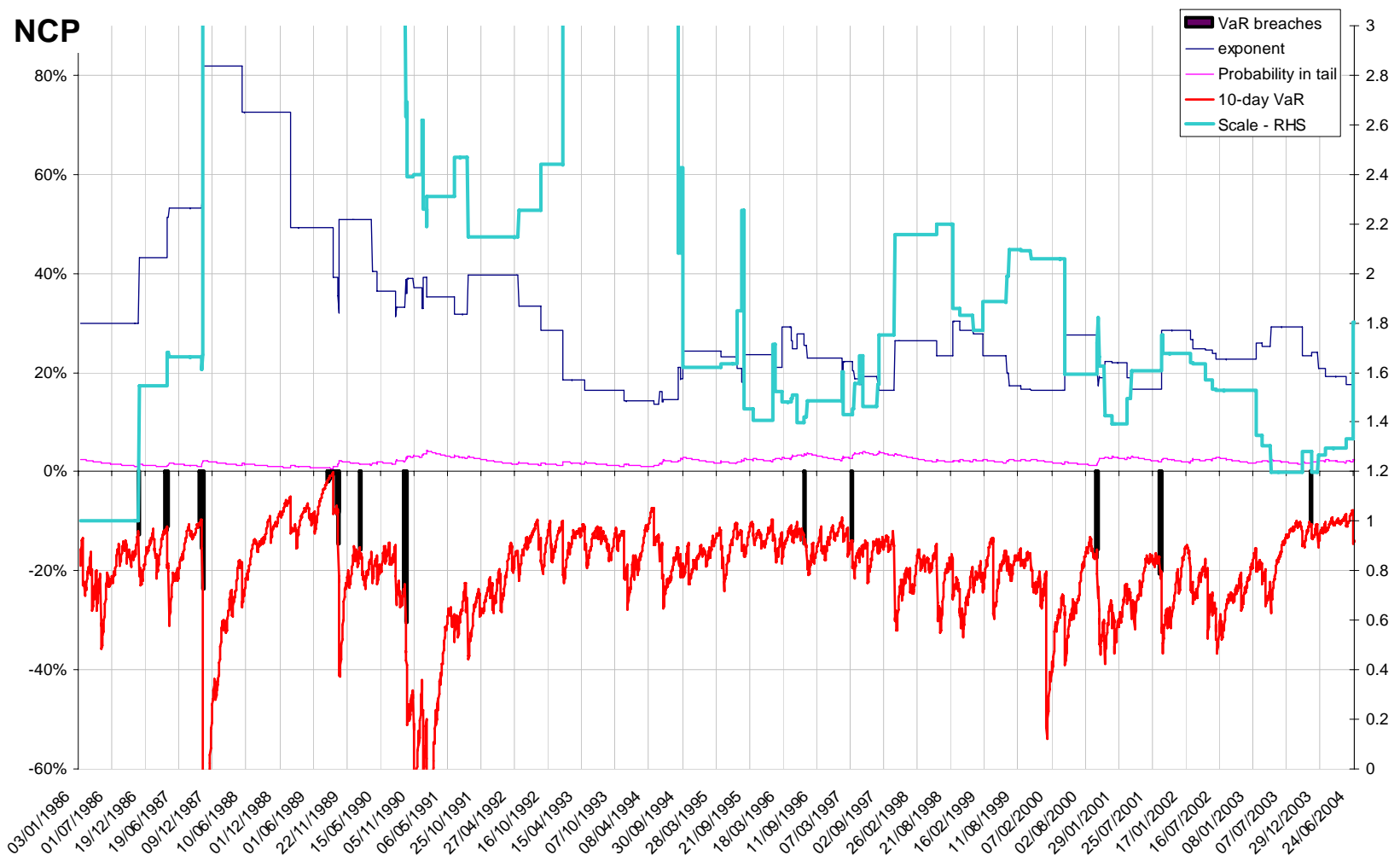
- $k = \Sigma \ln(x/u) / n$ usual formula
- $K_t = jK_{t-1} + (1-j)I_t \ln(x/u)$
- $k_t = K_t / p_t$
- shape parameter only changes when we get a tail value

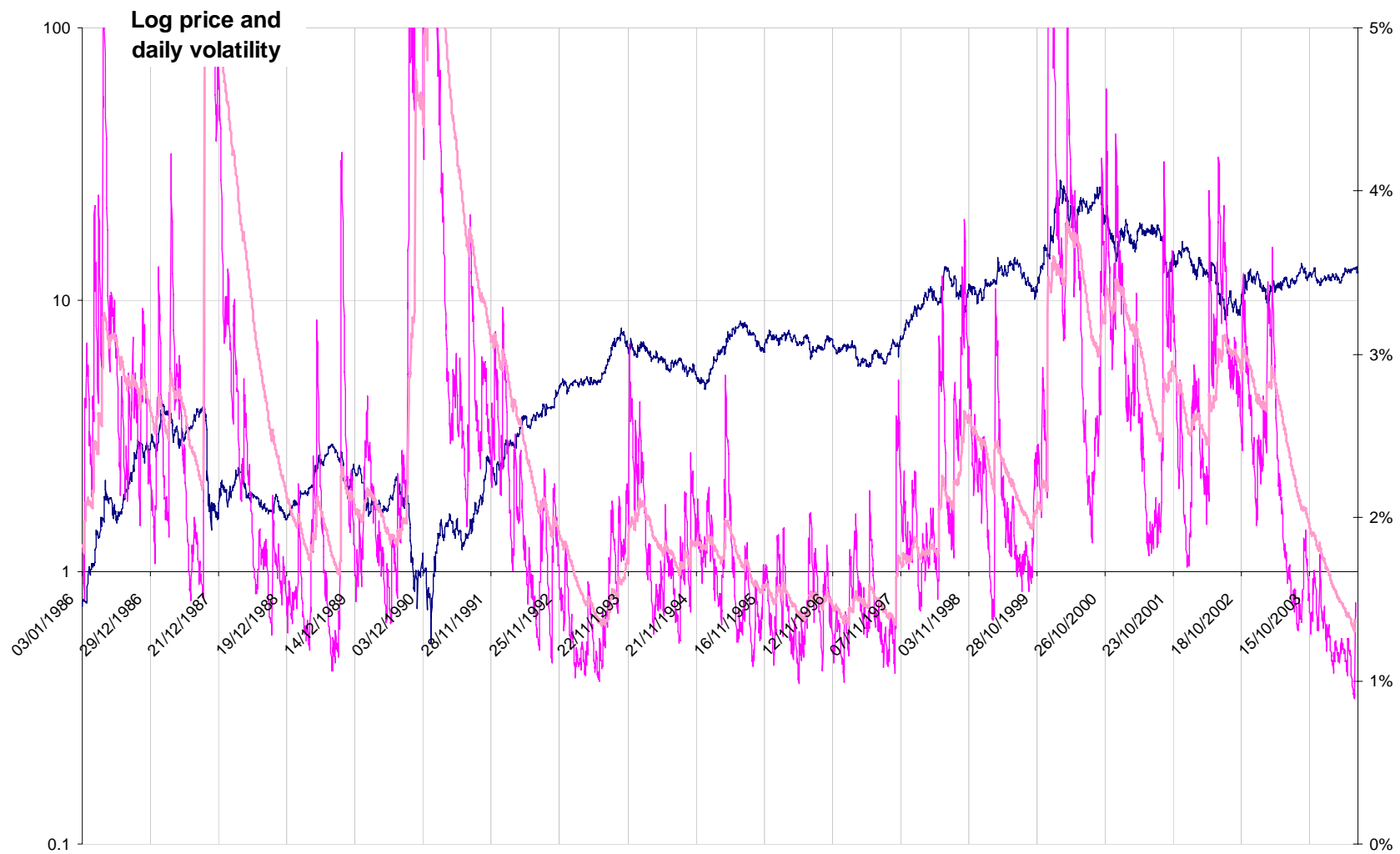
Scale parameter

- $S_t = \sum_{i=1}^n \ln(x/u)^{k(i)} (1-j)j^{(n-i)} I_i + s_0 j^n$
- $s_t = S_t / p_t$
- Scale parameter only changes when we're in the tail

EW-VaR

- $VaR_q = u + T_q$
- $T_q = (s/k) [(q/p)^{-k} - 1]$
- new value calculate at each point
- this can be put in spreadsheet





Observations

- VaR is inadequate when $p \approx q$
 - see late 1989
- Shape parameter is much more stable than scale
- VaR is variable

Uses

- Gives a feel for how EVT estimates are variable
- Possible explanatory power for company risk

Bibliography

- McNeil, Alexander, 1999, Extreme Value Theory for Risk Managers., for an introduction.
- Alexander, C. and E. Sheedy, 2007. *Model-Based Stress Tests: Linking Stress Tests to VaR for Market Risk* (ICMA Centre Discussion Paper in FinanceDP2007-02)