

Financial Risk 4-rd quarter 2016/17 Tuesdays 13.15 – 115.00 and Thursdays 15.15 – 17.00 in Euler

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"As an alternative to the traditional 30-year mortgage, we also offer an interest-only mortgage, balloon mortgage, reverse mortgage, upside down mortgage, inside out mortgage, loop-de-loop mortgage, and the spinning double axel mortgage with a triple lutz."



Gudrun January 2005 326 MEuro loss 72 % due to forest losses 4 times larger than second largest

This course:

- Learn some fun things about financial risks
- Learn some basic risk management tools from Extreme Value Statistics
- See some basic quantitative Credit Risk models

But not

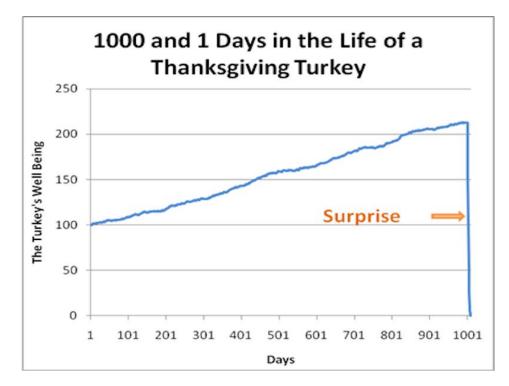
- A complete systematic account of financial risk management
- Financial time series modeling
- Black-Scholes option pricing methods
- Macroeconomics

and

There are risks which cannot be handled by mathematical models!!

Risk: event or action which prevents an institution from meeting its obligations or reaching its goals.

If one does not understand the real-world situation well enough, the best quantitative tools will not help. Taleb's Turkey example:



- Credit risk
- Market risk
- Operational risk
- Insurance risk
- Liquidity risk
- Reputational risk
- Legal risk
- and so on ...



Market risk: risk that the value of a portfolio changes due to changes of market prices, exchange rates etc.

Credit risk: risk that the value of a portfolio changes because a debtor cannot meet his obligations.

Operational risk: risk caused by problems in internal processes, people, systems

<u>Basel III</u>: s a global, voluntary regulatory framework on bank <u>capital adequacy</u>, <u>stress testing</u>, and <u>market liquidity risk</u>

<u>Solvency 2</u>: A <u>Directive</u> in <u>European Union law</u> that codifies and harmonises the EU insurance regulation. Primarily this concerns the amount of capital that <u>EU insurance</u> companies must hold to reduce the risk of <u>insolvency</u>.

Risk factors

L ="-P/L" = Loss – Profit = $f(X_1, ..., X_d)$, $X_1, ..., X_d$ risk factors, *e.g.* exchange rates, interest rates, index movements, stock prizes,

Example: Linear portfolio, α_i shares of stock *i*, stock prize $S_{t,i}$

$$L = -\sum_{i=1}^{d} \alpha_i S_{t+1,i} + \sum_{i=1}^{d} \alpha_i S_{t,i} = -\sum_{i=1}^{d} \alpha_i S_{t,i} \left(\frac{S_{t+1,i} - S_{t,i}}{S_{t,i}} \right)$$

How big is the risk?

Quantitative risk management methods:

- Historical data or historical simulation
- Stress testing ("scenarios")
- Sensitivity measures ("the greeks")
- Full statistical modeling (often multivariate normal + linear portfolio)
- Semiparametric modeling of the "tails" of the *loss-profit* distribution (univariate Extreme Value Statistics) (*this course*)
- Semiparametric modeling of the tails of the multivariate distribution of the risk factors (multivariate Extreme Value Statistics, "Copulas") + computation of the *loss-profit* distribution analytically or via stochastic simulation

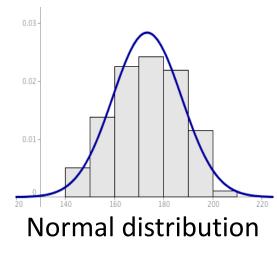
How big is the risk?

Matematics \rightarrow shapes of possible risk distributions

 \rightarrow choice of specific risk distribution (with uncertainty)

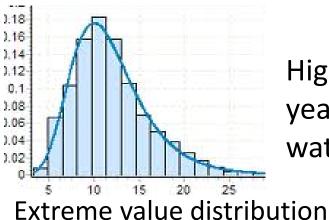
Statistics

Risk distribution \rightarrow quantifies risk, perhaps via VaR or ES, perhaps multivariate



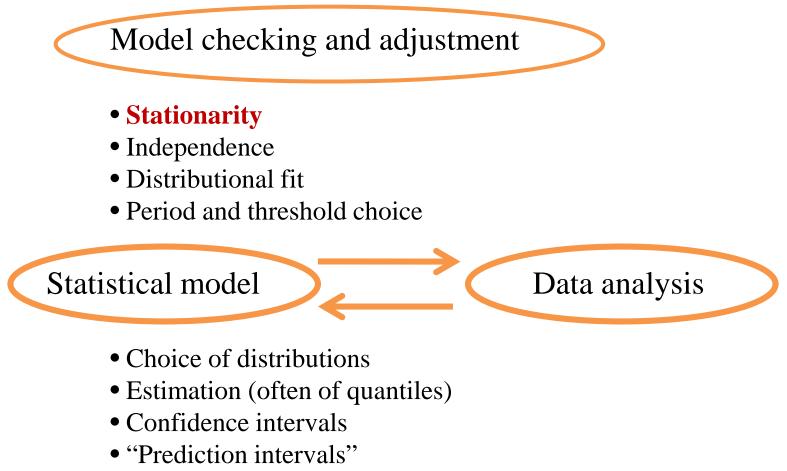
Error in the measured distance = sum of many small measurement errors





Highest water level during year = maximum of daily water levels





- Hypothesis testing
- Regression and dependence modeling (not this course)
- Prediction (*not this course*)
- Understanding!!

refresh your basic statistics knowledge!

Basic EVS:

--- Block Maxima: EV (= GEV) distribution for maxima --- Peaks over Thresholds: GP distribution for tails

Why?

- **stability:** maxima of variables which are EV distributed are also EV; going to higher levels preserves the GP distribution of exceedances (cf. "standard statistics: sums of normally distributed variables have a normal distribution)
- **asymptotics:** maxima of many independent variables are often (approximately) EV distributed; asymptotically tails are GP when maxima are EV (cf. "standard statistics: sums of many small "errors" are often (approximately) normally distributed the "central limit theorem")
- "transition": easy to go back and forth between GP and EV

but don't believe in models blindly

The Block Maxima method (Coles p. 45-53)

the Extreme Value (EV) distribution:

$$G(x) = \exp\{-(1 + \gamma \frac{x - \mu}{\sigma})_{+}^{-1/\gamma}\}$$

the special case $\gamma = 0$ is the Gumbel distribution

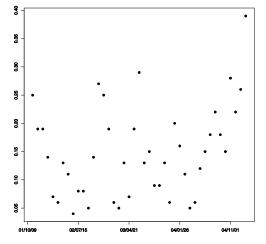
$$G_0(x) = \exp\{-\exp\{-\frac{x-\mu}{\sigma}\}\}$$

Fit to the measured block (=weekly, or monthly, or yearly, or ...) maxima, assuming independence, using maximum likelihood. Find *p*-th quantile from top (= value such that the risk that it is exceeded is 100p %) by solving

$$\hat{\bar{G}}(x_p) := 1 - \hat{G}(x_p) = p,$$

where ^ means that parameters are replaced by their estimated values.

confidence intervals via profile likelihood or delta method *(later lectures)*



Maximum long term US interest rate, constant maturity, nominal 1 month, percentage points

Some mathematics behind the Block Maxima Method:

 X_1, X_2, \ldots independent identically distributed random (i.i.d.) variables with distribution function (d.f.) F (*e.g.* daily interest rates, c.f. previous slide)

 $M_n = \max\{X_1, \dots, X_n\}$ = the maximum of the first *n* variables (*e.g. n* = 30 and *X* daily interest rates gives plot on previous slide)

$$P(M_n \le x) = P(X_1 \le x, \dots X_n \le x)$$

= $P(X_1 \le x) \times \dots \times P(X_n \le x) = F(x)^n$

Exercise: Show that the EV distributions are *max-stable*, *i.e.* that the maximum of *n* i.i.d. EV-distributed variables also have an EV distribution, *i.e.* that if $G(x) = \exp\{-(1 + \gamma \frac{x-\mu}{\sigma})^{-1/\gamma}_{+}\}$, then $G(x)^n = \exp\{-(1 + \gamma \frac{(x-\mu_n)/\sigma_n - \mu}{\sigma})^{-1/\gamma}\} = G(\frac{x-\mu_n}{\sigma_n})$

and find μ_n, σ_n .

Theorem: The distribution function G is max-stable if and only if it is an EV distribution

In the previous exercise it was shown that the EV distributions are maxstable, which proves half of this theorem. The other half consists of solving the functional equations

$$G(x)^n = G(\frac{x-\mu_n}{\sigma_n}), \text{ for } n = 1, 2, \dots$$

to find that the EV distributions are the only solutions.

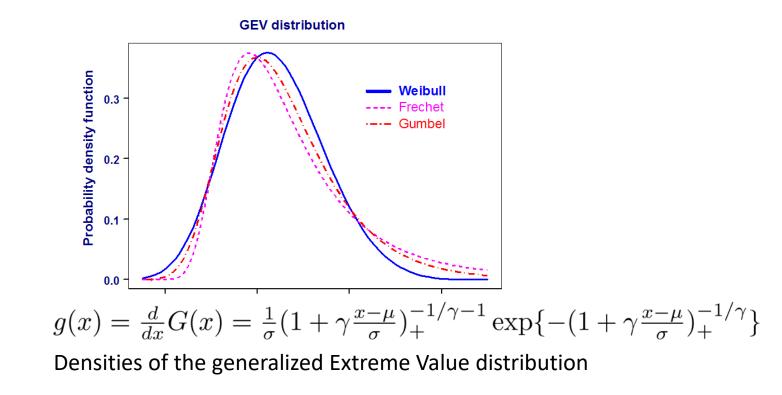
Theorem: If there are constants $b_n > 0, a_n$ such that

$$P\left(\frac{M_n - a_n}{b_n} \le x\right) \to G(x), \text{ as } n \to \infty \text{ for all } x$$

then G(x) is an EV distribution. (cf. the central limit theorem)

This is proved by showing that it follows that G(x) must be max-stable

Challenge: Prove this!



 $\gamma > 0$ Frechet distribution, finite left endpoint of distribution: $x > \mu + \frac{\sigma}{\gamma}$

 $\gamma = 0$ Gumbel distribution, unbounded support

 $\gamma < 0$ Weibull distribution, finite right endpoint of distribution: $x < \mu + \frac{\sigma}{|\gamma|}$

the distribution is "heavytailed" for $\gamma > 0$: then moments of order greater than $1/\gamma$ are infinite/don't exist; thus if $\gamma > 1/2$ then the variance doesn't exist, if $\gamma > 1$ then the mean doesn't exist either *Exercise:* For each of the following three distributions, show that, with the given norming constants a_n , b_n , there is a d.f. G(x) such that

$$P\left(\frac{M_n - a_n}{b_n} \le x\right) \to G(x)$$
, as $n \to \infty$ for all x ,

and find this G(x).

- 1.) The exponential distribution $F(x) = 1 e^{-\lambda x}, \ \lambda > 0, \ x \ge 0$ (use $a_n = \frac{\log n}{\lambda}, \ b_n = 1/\lambda$)
- 2.) The Pareto distribution $F(x) = 1 (\frac{\kappa}{\kappa + x})^{\alpha}, \quad \kappa, \alpha > 0, \quad x \ge 0$ (use $a_n = \kappa n^{1/\alpha} - \kappa, \quad b_n = \kappa n^{1/\alpha}/\alpha$)
- 3.) The uniform distribution F(x) = x, $0 \le x \le 1$ (use $a_n = 1 - 1/n$, $b_n = 1/n$)

(A perhaps unnecessary explanation) What does

$$P\left(\frac{M_n - a_n}{b_n} \le x\right) \to G(x), \text{ as } n \to \infty \text{ for all } x,$$

mean in practice? That $P(\frac{M_n-a_n}{b_n} \le x) \approx G(x)$, for large n, or,

with
$$y = b_n x + a_n$$
 and $G(x) = \exp\{-(1 + \gamma \frac{x - \mu'}{\sigma'}))^{-1/\gamma}\}$, that

$$P(M_n \le y) \approx G(\frac{y - a_n}{b_n}) = \exp\{-(1 + \gamma \frac{y - (a_n + b_n \mu')}{b_n \sigma'})^{-1/\gamma}\}$$

= $\exp\{-(1 + \gamma \frac{y - \mu}{\sigma})^{-1/\gamma}\}, \text{ for } \mu = a_n + b_n \mu', \ \sigma = b_n \sigma.$

Since all the parameters are unknown anyway, we are left with the problem of estimating μ , σ from data, *i.e.* with the Block Maxima method.