

SENSITIVITY ANALYSIS OF VALUES AT RISK

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Agenda

1. Motivation

- Risk Management / Portfolio Selection

2. Main Theoretical Results

- Analytical Expressions for 1st and 2nd Derivatives of Value at Risk (VaR) w.r.t. Portfolio Allocation
- Analytical Expressions of first derivatives of expected shortfall

3. Some Examples

- Gaussian Distribution with/without Unobserved Heterogeneity

4. Application

- VaR Efficient Portfolios

5. Estimation Procedure

- Nonparametric Approach (Kernel)

6. Empirical Illustration

- French Stock Data

1. Motivation

- Framework
 - regulatory environment
 - proprietary Risk Measurement Models
 - VaR = **Synthetic** Measure of Risk Portfolio

- *Risk Management*
 - risk analysis
 - risk control
 - portfolio optimization

- Need to go further than a single number

- What are the *risk drivers* ?
 1. VaR of subportfolios
 - ex-post analysis
 - aggregation issues
 - VaR is neither additive or subadditive
 2. Statistical analysis of *risk scenarios*
 3. Incremental VaR (**sensitivity analysis**)
 - ex-ante
 - management of risk limits
 - Large Portfolios preclude online computations
 - \implies **Sensitivity** of VaR w.r.t. portfolio allocation

- Why determine convexity of VaR w.r.t. portfolio allocation ?
- *Portfolio Selection under VaR Constraints*
 - Determination of Optimal Portfolios
 - Need to check that the set of portfolios satisfying VaR constraint is convex
- VaR may not be subadditive (i.e. a *coherent* measure of risk)
 - No bonus for diversification
 - Poor internal risk management
 - subadditivity can be statistically tested ...
 - and depends on financial intermediaries policies ...
 - * Credit risk,
 - * Out of the money options

2. Main Theoretical Results

- **VaR definition:**

- n Financial Assets with Prices $p_{i,t}$ at Date t

- Portfolio Value : $W_t(a) = \sum_{i=1}^n a_i p_{i,t} = a' p_t$

- $P_t [W_{t+1}(a) - W_t(a) + VaR_t(a, \alpha) < 0] = \alpha$

- Upper Quantile $(1 - \alpha)$...

- * since with $y_{t+1} = p_{t+1} - p_t$

- * $P_t [-a' y_{t+1} > VaR_t(a, \alpha)] = \alpha$

- **First properties**

- VaR is degree one positively homogeneous: $\lambda \geq 0$

- $VaR_t(\lambda a, \alpha) = \lambda \times VaR_t(a, \alpha)$

- $a' \frac{\partial VaR_t(a, \alpha)}{\partial a} = VaR_t(a, \alpha)$ (Euler)

- Risk contribution of i : $a_i \times \frac{\partial VaR_t(a, \alpha)}{\partial a_i}$

- VaR is **not always** subadditive:

- $VaR_t(a + b, \alpha) \leq VaR_t(a, \alpha) + VaR_t(b, \alpha)$?

- subadditivity + positive homogeneity \Rightarrow convexity of $VaR(a, \alpha)$ (w.r.t portfolio allocation a)

- **Sensitivities of VaR**

i) *1st derivative of VaR*

$$\frac{\partial VaR_t(a, \alpha)}{\partial a} = -E_t[y_{t+1} | a'y_{t+1} = -VaR_t(a, \alpha)]$$

ii) *2nd derivative of VaR*

$$\begin{aligned} \frac{\partial^2 VaR_t(a, \alpha)}{\partial a \partial a'} &= \frac{\partial \log g_{a,t}(-VaR_t(a, \alpha))}{\partial z} V_t[y_{t+1} | a'y_{t+1} = -VaR_t(a, \alpha)] \\ &\quad - \left\{ \frac{\partial}{\partial z} V_t[y_{t+1} | a'y_{t+1} = -z] \right\}_{z=VaR_t(a, \alpha)}, \end{aligned}$$

with $g_{a,t}$ the conditional p.d.f of $a'y_{t+1}$

• **Expected Shortfall definition:**

– $m_t(a, \alpha) = E_t[-a'y_{t+1} \mid -a'y_{t+1} > VaR_t(a, \alpha)]$

– Also known as *Mean Excess Loss*, as *TailVaR* or as *Lower Partial Moment*.

– Another commonly used measure of risk

– Expected Shortfall is subadditive

i) *1st derivative of Expected Shortfall*

$$\frac{\partial m_t(a, \alpha)}{\partial a} = -E_t[y_{t+1} \mid -a'y_{t+1} > VaR_t(a, \alpha)]$$

Similar result to that obtained on sensitivity of VaR

3. Some Examples

⇒ First example: Gaussian Distribution

- $y_{t+1} \sim N(\mu_t, \Omega_t)$
- $VaR_t(a, \alpha) = -a'\mu_t + (a'\Omega_t a)^{1/2} z_{1-\alpha}$
- $z_{1-\alpha}$ quantile of level $1 - \alpha$ of Gaussian distribution.

Remark:

- $(a'\Omega_t a)^{1/2}$: standard dev. of portfolio absolute returns
- subadditivity of standard deviations:
- $((a + b)'\Omega_t(a + b))^{1/2} \leq (a'\Omega_t a)^{1/2} + (b'\Omega_t b)^{1/2}$
- ⇒ In the Gaussian case, VaR is always subadditive

i) 1st derivative of VaR

$$\begin{aligned}\frac{\partial VaR_t(a, \alpha)}{\partial a} &= -\mu_t + \frac{\Omega_t a}{(a'\Omega_t a)^{1/2}} z_{1-\alpha} \\ &= -\mu_t + \frac{\Omega_t a}{a'\Omega_t a} (VaR_t(a, \alpha) + a'\mu_t) \\ &= -E_t [y_{t+1} | a'y_{t+1} = -VaR_t(a, \alpha)]\end{aligned}$$

ii) 2nd derivative of VaR

$$\frac{\partial^2 VaR_t(a, \alpha)}{\partial a \partial a'} = \frac{z_{1-\alpha}}{(a' \Omega_t a)^{1/2}} \left[\Omega_t - \frac{\Omega_t a a' \Omega_t}{a' \Omega_t a} \right]$$

$$\frac{\partial^2 VaR_t(a, \alpha)}{\partial a \partial a'} = \frac{z_{1-\alpha}}{(a' \Omega_t a)^{1/2}} V_t[y_{t+1} | a' y_{t+1} = -VaR_t(a, \alpha)]$$

Remark :

$$\frac{\partial \log g_{a,t}(-VaR_t(a, \alpha))}{\partial z} = \frac{VaR_t(a, \alpha) + a' \mu_t}{a' \Omega_t a} = \frac{z_{1-\alpha}}{(a' \Omega_t a)^{1/2}}$$

2nd term = 0 (conditional homoscedasticity)

- Expected Shortfall $m_t(a, \alpha)$
- $m_t(a, \alpha) = -a' \mu_t + (a' \Omega_t a)^{1/2} \frac{\varphi(z_{1-\alpha})}{\alpha}$
 – φ : Gaussian density

i) 1st derivative of Expected Shortfall

$$\frac{\partial m_t(a, \alpha)}{\partial a} = -\mu_t + \frac{\Omega_t a}{(a' \Omega_t a)^{1/2}} \frac{\varphi(z_{1-\alpha})}{\alpha}$$

⇒ Second example:

Gaussian with *Unobserved Heterogeneity*

- $y_{t+1} \mid u \sim N(0, \Omega_t(u))$
- with heterogeneity factor u with distribution Π
- Gaussian Random Walk with Stochastic Volatility

Check of VaR Convexity:

$$\frac{\partial \log g_{a,t}(-VaR_t(a, \alpha))}{\partial z} = VaR_t(a, \alpha) E_{\tilde{\Pi}} \left[\frac{1}{a' \Omega_t(u) a} \right] > 0.$$

$$\begin{aligned} -\frac{\partial}{\partial z} V_t[y_{t+1} \mid a' y_{t+1} = -z] &= -\frac{\partial}{\partial z} \left[V_{\Pi} \left[-z \frac{\Omega_t(u) a}{a' \Omega_t(u) a} \right] \right] \\ &= +2z V_{\Pi} \left[\frac{\Omega_t(u) a}{a' \Omega_t(u) a} \right], \end{aligned}$$

which is nonnegative for $z = VaR_t(a, \alpha)$!

4. Application : VaR Efficient Portfolios

Budget w allocated among

n Risky Assets and 1 Riskfree Asset (r)

$$\begin{cases} \max_a & a' E_t y_{t+1} \\ \text{s.t.} & VaR_t(a; \alpha) \leq VaR_o - w(1 + r) = V\tilde{a}R_o \end{cases}$$

$VaR_o =$ Bound for Authorized Risk (CAD)

First Order Conditions :

$$\begin{cases} E_t y_{t+1} = -\lambda_t^* \frac{\partial VaR_t}{\partial a}(a_t^*, \alpha) \\ VaR_t(a_t^*, \alpha) = V\tilde{a}R_o \end{cases}$$

Proportionality between Global and Local Expectations :

$$E_t y_{t+1} = \lambda_t^* E_t \left[y_{t+1} | a_t^* \right] y_{t+1} = -V\tilde{a}R_o$$

5. Estimation Procedure

Nonparametric approach (kernel) : i.i.d. returns

- Estimation of VaR

$$P[-a'y_{t+1} > VaR(a, \alpha)] = \alpha$$

estimated by Gaussian Kernel

$$\frac{1}{T} \sum_{t=1}^T \Phi \left(\frac{-a'y_t - \widehat{VaR}}{h} \right) = \alpha$$

Gauss-Newton Algorithm :

$$var^{(p+1)} = var^{(p)} + \frac{\frac{1}{T} \sum_{t=1}^T \Phi \left(\frac{-a'y_t - var^{(p)}}{h} \right) - \alpha}{\frac{1}{Th} \sum_{t=1}^T \varphi \left(\frac{a'y_t + var^{(p)}}{h} \right)}$$

Starting values : Gaussian VaR or Empirical Quantile

- Convexity of VaR

Hessian $\frac{\partial^2 VaR(a, \alpha)}{\partial a \partial a'}$ positive semidefinite

If, for negative z values

$$\frac{\partial \log g_{a,t}(z)}{\partial z} > 0 \text{ and } \frac{\partial V[y_{t+1} | a'y_{t+1} = z]}{\partial z} \gg 0$$

Estimator of p.d.f. of portfolio value

$$\hat{g}_a(z) = \frac{1}{Th} \sum_{t=1}^T \varphi \left(\frac{a'y_t - z}{h} \right)$$

Estimator of conditional variance

$$\hat{V}[y_{t+1} | a'y_{t+1} = z] = \frac{\sum_{t=1}^T y_t y_t' \varphi \left(\frac{a'y_t - z}{h} \right)}{\sum_{t=1}^T \varphi \left(\frac{a'y_t - z}{h} \right)} - \frac{\sum_{t=1}^T y_t \varphi \left(\frac{a'y_t - z}{h} \right) \sum_{t=1}^T y_t' \varphi \left(\frac{a'y_t - z}{h} \right)}{\left[\sum_{t=1}^T \varphi \left(\frac{a'y_t - z}{h} \right) \right]^2}$$

- Estimation of VaR efficient portfolio

Simple forms of 1st and 2nd derivatives of VaR

⇒ Gauss-Newton Algorithm

$$a^{(p+1)} = a^{(p)} - \left[\frac{\partial^2 VaR}{\partial a \partial a'}(a^{(p)}, \alpha) \right]^{-1} \frac{\partial VaR}{\partial a}(a^{(p)}, \alpha) + \left[\frac{2(V\bar{a}R_o - VaR(a^{(p)}, \alpha)) + Q(a^{(p)}, \alpha)}{Ey'_{t+1} \left[\frac{\partial^2 VaR}{\partial a \partial a'}(a^{(p)}, \alpha) \right]^{-1} Ey_{t+1}} \right]^{1/2} \times \left[\frac{\partial^2 VaR}{\partial a \partial a'}(a^{(p)}, \alpha) \right]^{-1} Ey_{t+1}$$

with

$$Q(a^{(p)}, \alpha) = \frac{\partial VaR}{\partial a'}(a^{(p)}, \alpha) \left[\frac{\partial^2 VaR}{\partial a \partial a'}(a^{(p)}, \alpha) \right]^{-1} \frac{\partial VaR}{\partial a}(a^{(p)}, \alpha)$$

Theoretical recursion replaced by empirical counterpart

6. Empirical Illustration

French Stock Data from CAC 40

- Thompson-CSF (electronic devices)
- L'Oréal (cosmetics)

Daily Returns : 04/01/1997 to 05/04/1999 (546 obs.)

Empirical Results :

- * Standard Normal VaR underestimate (Skew. and Kurt.)
- * Smoother patterns for Kernel estimates
- * Nonmonotonicity of Sensitivities
- * Check for Convexity fails for some allocations
- * VaR Symmetry lost
- * VaR Efficient Portfolios = Tangency Points of

$$a_1\hat{\mu}_1 + a_2\hat{\mu}_2 = cst \text{ with IsoVar curve of level } \widetilde{VaR}_o$$