

12.1

## VALUE AT RISK

VALUE AT RISK VaR IS THE MAX AMOUNT YOU MAY LOSE WITH A CERTAIN PROBABILITY.

$V_0$  INITIAL PORTFOLIO VALUE

$V_1$  FINAL PORTFOLIO VALUE

$$P(V_1 - V_0 \leq -\text{VaR}) = c \quad \text{WHERE } 1-c \text{ IS THE PROBABILITY TO LOSE NOT MORE THAN VaR}$$

$\text{VaR}(c)$  IS THE  $c$ -LEVEL VALUE AT RISK

BE  $G$  THE PRESENT VALUE OF A GAIN FROM AN INVESTMENT.

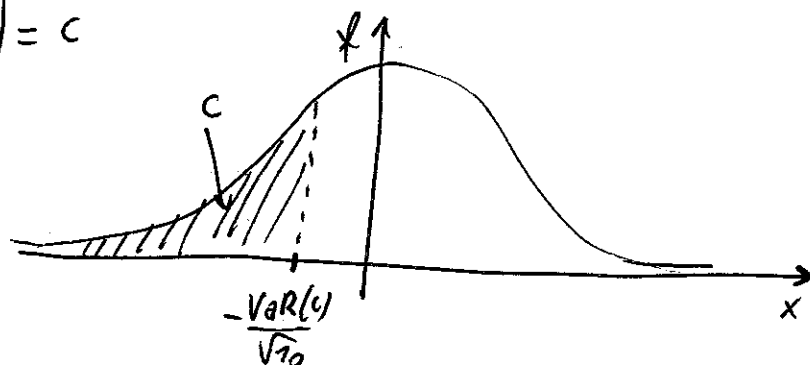
$$P(G < -\text{VaR}(c)) = c \quad \text{DETERMINATES VaR AS FUNCTION OF } c$$

EXAMPLE:  $G \sim N(0, \sigma^2)$

$$P(G < -\text{VaR}(c)) = \Phi\left(\frac{-\text{VaR}(c) - 0}{\sigma}\right) = \Phi\left(-\frac{\text{VaR}(c)}{\sigma}\right)$$

FOR EXAMPLE, IF  $\sigma^2 = 10$  AND IF WE WANT  $c = 10\%$

$$\Phi\left(-\frac{\text{VaR}(c)}{\sqrt{10}}\right) = c$$



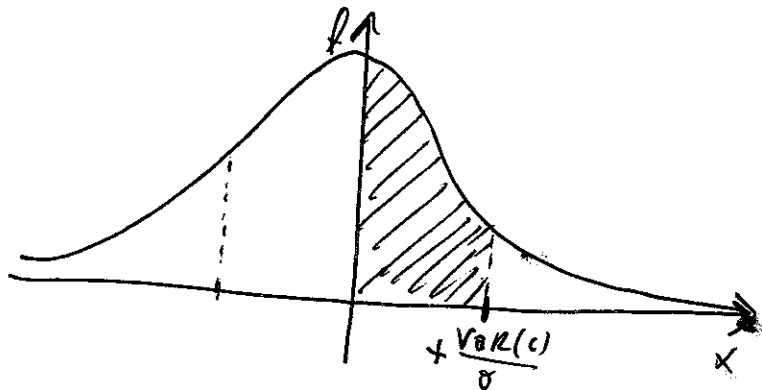
THIS AREA CAN BE FOUND ON TABLES. LOOK FOR AN AREA OF 0.1 AND YOU WILL FIND A VALUE OF -1.28

12.2

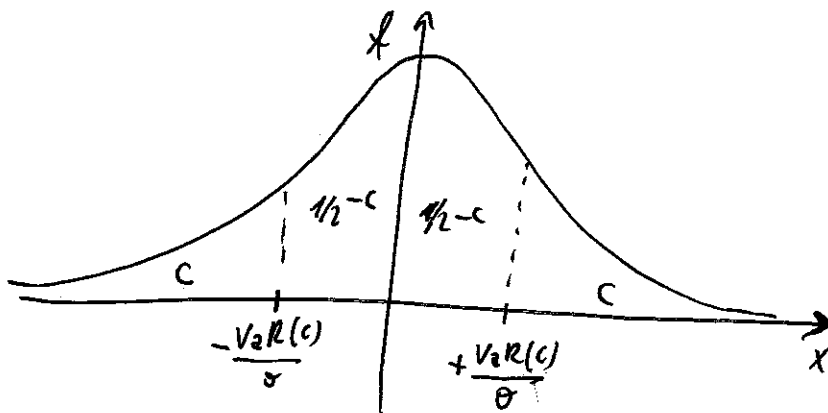
$$\frac{-\text{Var}(c)}{\sigma} = -1,28$$

$$\text{Var}(c) \approx 4,05 \text{ for } c=10\%$$

NOTE HOWEVER THAT OFTEN TABLES GIVE THIS AREA



AND TO FIND THE CORRECT VALUES IN THESE TABLES, YOU SHOULD LOOK FOR AN AREA WITH VALUE  $\frac{1}{2} - c$  (SINCE GAUSSIAN IS SYMMETRIC) AND REVERSE THE RESULT'S SIGN.



C-LEVEL CONDITIONAL VALUE AT RISK  $\text{CVAR}(c)$  IS

$$-\frac{1}{c} \int_{-\infty}^{-\text{Var}(c)} s \cdot f(s) ds$$

IT IS BASICALLY THE EXPECTED VALUE OF THE GAIN, BUT ONLY FROM  $-\infty$  TO  $-\text{Var}(c)$ .  $\frac{1}{c}$  IS A SCALE FACTOR, BECAUSE WHEN  $c$  IS VERY SMALL  $-\text{Var}(c)$  BECOMES VERY SMALL AND THE INTEGRAL WOULD BE, WITHOUT  $\frac{1}{c}$ , EXTREMELY SMALL.

NOTE THAT IT WORKS ONLY FOR CONTINUOUS R.V.

12.3

ALTERNATIVE EXPRESSION FOR CVAR

INDICATOR FUNCTION:  $X_{\{Y < \alpha\}} = \begin{cases} 1 & \text{when } Y < \alpha \\ 0 & \text{when } Y \geq \alpha \end{cases}$

THEREFORE

$$\text{CVAR}(c) = -\frac{1}{c} \int_{-\infty}^{-\text{Var}(c)} s \cdot f(s) ds = -\frac{1}{c} \int_{-\infty}^{+\infty} s \cdot f(s) \cdot X_{\{s < -\text{Var}(c)\}} ds =$$

$$= -\frac{1}{c} \cdot E(G \cdot X_{\{G < -\text{Var}(c)\}}) = -\frac{E(G \cdot X_{\{G < -\text{Var}(c)\}})}{P(G < -\text{Var}(c))}$$

THIS FORMULA WORKS ALSO FOR DISCRETE R.V.

ANOTHER ALTERNATIVE EXPRESSION FOR CVAR

$F_G(x | G < -\text{Var}(c)) = P(G < x | G < -\text{Var}(c)) =$  IS THE  
CONDITIONAL CUMULATIVE DISTRIBUTION FUNCTION

$$= P((G < x) \cap (G < -\text{Var}(c))) / P(G < -\text{Var}(c)) =$$

$$= \frac{P(G < \min(x, -\text{Var}(c)))}{P(G < -\text{Var}(c))}$$

$$F_G(x | G < -\text{Var}(c)) = \begin{cases} x \leq -\text{Var}(c) & P(G < x) / P(G < -\text{Var}(c)) \\ x \geq -\text{Var}(c) & P(G < -\text{Var}(c)) / P(G < -\text{Var}(c)) = 1 \end{cases}$$

THE CORRESPONDING DENSITY FUNCTION

$$f_G(x | G < -\text{Var}(c)) = \begin{cases} x \leq -\text{Var}(c) & f(x) / P(G < -\text{Var}(c)) \\ x > -\text{Var}(c) & 0 \end{cases}$$

$$12.4) \quad C\text{VAR}(c) = -\frac{1}{c} \int_{-\infty}^{-\text{VAR}(c)} s \cdot f(s) ds = - \int_{-\infty}^{-\text{VAR}(c)} s \cdot \frac{f(s)}{P(G < -\text{VAR})} ds$$

$$= - \int_{-\infty}^{+\infty} s \cdot f_G(s | G < -\text{VAR}) ds = -E(G | G < -\text{VAR}(c))$$

THIS FORMULA WORKS ALSO FOR DISCRETE R.V.

EXAMPLE WITH THE NORMAL DISTRIBUTION

$G \sim N(\mu, \sigma^2)$  FIND  $C\text{VAR}(c)$  AS A FUNCTION OF  $c, \mu, \sigma^2, \text{VAR}$

$$c = \int_{-\infty}^{-\text{VAR}(c)} f_N(s) ds = \Phi\left(\frac{-\text{VAR}(c) - \mu}{\sigma}\right) \quad \text{FROM HERE VAR CAN BE CALCULATED}$$

$$C\text{VAR}(c) = -\frac{1}{c} \int_{-\infty}^{-\text{VAR}} s \cdot \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(s-\mu)^2}{2\sigma^2}} ds = -\frac{1}{c} \int_{-\infty}^{-\text{VAR}} (s-\mu) \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(s-\mu)^2}{2\sigma^2}} ds +$$

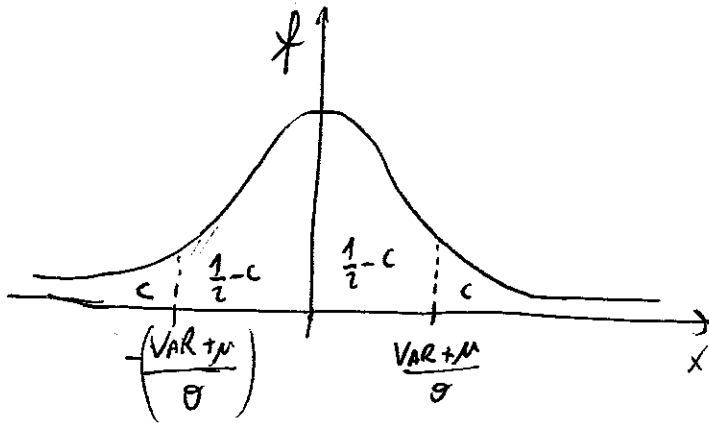
$$-\frac{1}{c} \int_{-\infty}^{-\text{VAR}} \mu \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(s-\mu)^2}{2\sigma^2}} ds = \left[ -\frac{1}{c} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(s-\mu)^2}{2\sigma^2}} \cdot \left(-\frac{2\sigma^2}{2}\right) \right]_{-\infty}^{-\text{VAR}} +$$

$$-\frac{1}{c} \int_{-\infty}^{-\text{VAR}} \mu f_{N(\mu, \sigma^2)}(s) ds = +\frac{1}{c} \frac{\sigma}{\sqrt{2\pi}} e^{-\frac{(-\text{VAR}-\mu)^2}{2\sigma^2}} - 0 - \frac{\mu}{c} \cdot c =$$

$$= -\mu + \frac{\sigma}{c\sqrt{2\pi}} e^{-\frac{(\text{VAR}+\mu)^2}{2\sigma^2}}$$

THIS FORMULA IS VALID FOR ANY  $G \sim N(\mu, \sigma^2)$

12.5)  $G \sim N(\mu, \sigma^2)$   $c = 0,01$  FIND CVAR AS FUNCTION OF  $\mu, \sigma^2$



$$0.01 = \phi\left(\frac{-VAR - \mu}{\sigma}\right)$$

$$\frac{1}{2} - c = \frac{1}{2} - 0,01 = 0,49$$

$$-\frac{VAR + \mu}{\sigma} = -2,33$$

$$VAR(0,01) = -\mu + 2,33\sigma$$

$$CVAR(0,01) = -\mu + \frac{\sigma}{0,01\sqrt{\pi^2}} e^{-\frac{(\mu + 2,33\sigma - \mu)^2}{2\sigma^2}} =$$

$$\approx -\mu + 39,89\sigma e^{-\frac{2,33^2}{2}} \approx -\mu + 2,64\sigma$$

IT IS DIFFERENT FROM VAR.

12.6

SUPPOSE  $G = \Delta V$  IS  $V(-5; 5)$   
 FIND  $\text{VAR}(0,1)$  AND  $\text{CVAR}(0,1)$

$$0,1 = \int_{-\infty}^{-\text{VAR}(0,1)} f_V(s) ds = \int_{-5}^{-\text{VAR}(0,1)} \frac{1}{10} ds = \left[ \frac{s}{10} \right]_{-5}^{-\text{VAR}(0,1)} = \frac{-\text{VAR}(0,1) + 5}{10}$$

$$1 = -\text{VAR}(0,1) + 5$$

$$\text{VAR}(0,1) = 4$$

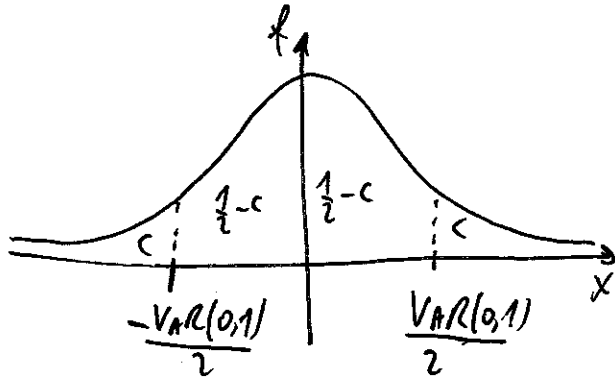
$$\text{CVAR}(0,1) = -\frac{1}{0,1} \int_{-\infty}^{-\text{VAR}(0,1)} s \cdot f_V(s) ds = -\frac{1}{0,1} \int_{-5}^{-4} s \cdot \frac{1}{10} ds = -\left[ \frac{s^2}{2} \right]_{-5}^{-4} = -\frac{16 - 25}{2} = +4,5$$

D

12.7

SUPPOSE  $G = \Delta V$  IS  $N(0; 2)$   $\mu = 0$   $\sigma^2 = 2$   
 FIND  $VAR(0,1)$  AND  $CVAR(0,1)$

$$0,1 = \int_{-\infty}^{-VAR(0,1)} \frac{1}{\sqrt{2\pi \cdot 2}} e^{-\frac{s^2}{2 \cdot 2}} ds = \Phi\left(\frac{-VAR(0,1) - 0}{\sqrt{2}}\right)$$



$$\frac{1}{2} - c = \frac{1}{2} - 0,1 = 0,4$$

$$\frac{-VAR(0,1)}{\sqrt{2}} = -1,285$$

$$VAR(0,1) = 1,81726$$

$$CVAR(0,1) = -\frac{1}{0,1} \int_{-\infty}^{-1,82} s \cdot \frac{1}{\sqrt{2\pi \cdot 2}} e^{-\frac{s^2}{2 \cdot 2}} ds$$

WE CAN USE THIS FORMULA, WHICH IS QUITE COMPLICATED OR APPLY IMMEDIATELY THE FORMULA FOR THE NORMAL

$$CVAR(c) = -\mu + \frac{\sigma}{c\sqrt{2\pi}} e^{-\frac{(\mu + VAR(c))^2}{2\sigma^2}}$$

THE FORMULA LEADS TO  $CVAR(0,1) = -0 + \frac{\sqrt{2}}{0,1 \cdot \sqrt{2\pi}} e^{-\frac{1,82^2}{4}} = 2,4716$

ON THE OTHER HAND, DOING ALL THE CALCULATIONS

$$CVAR(0,1) = -2,82 \int_{-\infty}^{-1,82} s \cdot e^{-\frac{s^2}{4}} ds = +2,82 \left[ e^{-\frac{s^2}{4}} \cdot 2 \right]_{-\infty}^{-1,82} = 5,64 \cdot \left( e^{-\frac{1,82^2}{4}} - 0 \right) = 2,4702$$

□

12.8

$CVAR(0.1) = 3x + 2y$  WHERE  $(x, y)$  MAY BE

- $(1, 2)$
- $(2, 1)$
- $(0.5, 5)$
- $(0.25, 3)$

WHICH INVESTMENT IS THE BEST?

CVAR IS THE EXPECTED VALUE OF THE LOSS FROM  $-\infty$  TO  $VAR(0.1)$ .

THEREFORE WE WOULD LIKE THAT IT BE AS SMALL AS POSSIBLE

$(1, 2) \rightarrow 3 + 4 = 7$

$(2, 1) \rightarrow 6 + 2 = 8$

$(0.5, 5) \rightarrow 1.5 + 10 = 11.5$

$(0.25, 3) \rightarrow 0.75 + 6 = 6.75 \rightarrow$  THIS IS THE BEST INVESTMENT

SO THE BEST INVESTMENT IS ALWAYS THE ONE WITH THE LOWEST VAR OR CVAR (WHEN C IS THE SAME)

□