

# Mathematical Tools for Asset Management

## MTH6134

### Measures of Investment Risk

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- ▶ Commonly Used Measures of Investment Risks
  - ▶ Expected Value of Returns
  - ▶ Variance of Returns
  - ▶ Semi-Variance of Returns
  - ▶ Shortfall Probabilities
  - ▶ Value at Risk
  - ▶ Expected Shortfall

# Measures of Investment Risk

Question: How can we rank investments/gambles/lotteries if

- ▶ if **we don't know the whole distribution of returns of investment/asset**

Answer: Use **partial known information on the distribution of returns** (i.e. moments of distribution)

**Return on Asset:** percentage increase in the market value of an asset over a specified period

- ▶ Discrete Random Variable  $X$  can take values  $x_1, \dots, x_n$  with probabilities  $p_1, \dots, p_n$  and  $\sum_i p_i = 1$
- ▶ Continuous Random Variable  $X$  can take values across a range characterised by a p.d.f.  $f(x)$

# Measures of Investment Risk

- ▶ Expected Value/Mean

$$E(X) \equiv \mu = \sum_i p_i x_i \text{ if } X \text{ is discrete}$$

$$E(X) \equiv \mu = \int_{-\infty}^{\infty} x f(x) dx \text{ if } X \text{ is continuous}$$

- ▶ Measures of investment risk:
  - ▶ variance of returns
  - ▶ downside semi-variance of return
  - ▶ shortfall probabilities
  - ▶ value at risk/tail value at risk

# Variance of Returns

Most theories of investment risk use variance of return as the measure of risk

$$\text{Var}(X) \equiv \sigma^2 = \sum_i (x_i - \mu)^2 p_i \text{ if } X \text{ is discrete}$$

$$\text{Var}(X) \equiv \sigma^2 = \int_{-\infty}^{\infty} (x - \mu)^2 f(x) dx \text{ if } X \text{ is continuous}$$

# Variance of Returns

The variance of returns  $\text{Var}(R) = \mathbb{E}((R - \mathbb{E}(R))^2)$

- measures uncertainty in terms of scatter around the expectation,
- measures distance between realised and expected return  $R - \mathbb{E}(R)$ ,
- by the square the sign vanishes and larger deviations are weighted higher than smaller ones,
- by taking the outer expectation, the deviations are weighted according to their likelihoods.
- The variance is 0 if there is no risk!

## Example

The investment annual returns  $X$  for a particular stock are modelled using a pdf:

$$f(x) = 750 \left( 0.01 - (x - 0.05)^2 \right)$$
$$-0.05 \leq x \leq 0.15 \text{ or } -5\% \leq x \leq 15\%$$

Verify that  $f(x)$  is a proper pdf:

$$\begin{aligned} \int_{-0.05}^{0.15} f(x) dx &= \int_{-0.05}^{0.15} 750 \left( 0.01 - (x - 0.05)^2 \right) dx \\ &= 750 \int_{-0.05}^{0.15} (0.0075 + 0.1x - x^2) dx \\ &= 750 \left[ 0.0075x + \frac{0.1x^2}{2} - \frac{x^3}{3} \right]_{-0.05}^{0.15} \\ &= 1 \end{aligned}$$



# Example

The average of the returns:

$$\begin{aligned} E(X) &= 750 \int_{-0.05}^{0.15} x \left( 0.01 - (x - 0.05)^2 \right) dx \\ &= 750 \int_{-0.05}^{0.15} x \left( 0.01 - (x^2 - 0.1x + 0.0025) \right) dx \\ &= 750 \int_{-0.05}^{0.15} (0.0075x + 0.1x^2 - x^3) dx \\ &= 750 \left[ \frac{0.0075}{2} x^2 + \frac{0.1x^3}{3} - \frac{x^4}{4} \right]_{-0.05}^{0.15} = 0.05 \end{aligned}$$

## Example

The variance of the returns for the same stock:

$$\begin{aligned} \text{Var}(X) &= \int_{-0.05}^{0.15} 750 \left( 0.01 - (x - 0.05)^2 \right) (x - 0.05)^2 dx \\ &= 750 \int_{-0.05}^{0.15} \left( 0.01 (x - 0.05)^2 - (x - 0.05)^4 \right) dx \\ &= 750 \left[ \frac{0.01}{3} (x - 0.05)^3 - \frac{1}{5} (x - 0.05)^5 \right]_{-0.05}^{0.15} = 0.002 \end{aligned}$$

# Variance of Returns - Further Examples

Consider two assets with today's price £1,000 and the following distribution for the price after 1 week:

	25%	25%	50%
Asset 1	£ 750	£1,000	£ 1,175
Asset 2	£ 550	£1,000	£ 1,275

- What is the expected return?
- What is the variance?
- Which asset would you prefer to reduce your risk?

$$\mathbb{E}(R_1) = -0.25 * 0.25 + 0 * 0.25 + 0.175 * 0.5 = 0.0250$$

$$\mathbb{E}(R_2) = -0.45 * 0.25 + 0 * 0.25 + 0.275 * 0.5 = 0.0250$$

$$\text{Var}(R_1) = \mathbb{E}[R_1^2] - \mathbb{E}[R_1]^2 = 0.0303$$

$$\text{Var}(R_2) = \mathbb{E}[R_2^2] - \mathbb{E}[R_2]^2 = 0.0878$$

**The variance of Asset 1 is smaller, and the expected returns are equal.  
Asset 1 with the smaller variance is preferred.**

# Variance of Returns

Consider two assets with today's price £1,000 and the following distribution for the price after 1 week:

	25%	25%	50%
Asset 1	£ 750	£1,000	£ 1,175
Asset 2	£ 800	£1,000	£ 1,500

- What is the expected return?
- What is the variance?
- Which asset would you prefer to reduce your risk?

$$\mathbb{E}(R_1) = 0.025, \quad \mathbb{E}(R_2) = 0.2$$

$$\text{Var}(R_1) = 0.0303, \quad \text{Var}(R_2) = 0.095$$

**The variance of Asset 2 is higher, but its expected return as well.**

# Variance of Returns

Today's price  $S_0$ : £1,000

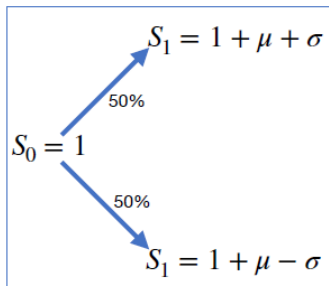
Tomorrow's price  $S_1$ , see table

	25%	25%	50%	$\mathbb{E}(R_0)$	$\text{Var}(R_0)$
Asset 1	£ 750	£1,000	£ 1,175	0.025	0.0303
Asset 2	£ 550	£1,000	£ 1,275	0.025	0.0878
Asset 3	£ 800	£1,000	£ 1,500	0.2	0.0950

For an investment decision consult at least both the returns'

- variance, and
- return!

# Variance of Returns: A binomial model



Expectation and variance:

$$\mathbb{E}(S_1) = 1 + \mu$$

$$\text{Var}(S_1) = \sigma^2$$

# Variance of Returns: A binomial Model

Examples:

$S_0 = 1$	50%	50%
$S_1$	$1 + \mu - \sigma$	$1 + \mu + \sigma$
$R_0$	$\mu - \sigma$	$\mu + \sigma$

$$R_0 = S_1/S_0 - 1$$

	unlucky (50%)	lucky (50%)	$\mathbb{E}(R_0)$	$\text{Var}(R_0)$
1) $\mu = 0.1, \sigma = 0$	1.1	1.1	0.1	0
2) $\mu = 0.1, \sigma = 1$	0.1	2.1	0.1	1
3) $\mu = 0.07, \sigma = 0$	1.07	1.07	0.07	0
4) $\mu = 0.2, \sigma = 0.2$	1.0	1.4	0.2	0.04
5) $\mu = 0.15, \sigma = 0.1$	1.05	1.25	0.15	0.01

# Variance of Returns: A binomial Model

Pairwise comparison

$S_0 = 1$	50%	50%
$R_0$	$\mu - \sigma$	$\mu + \sigma$

	unlucky (50%)	lucky (50%)	$\mathbb{E}(R_0)$	$\text{Var}(R_0)$
1) $\mu = 0.1, \sigma = 0$	1.1	1.1	0.1	0
2) $\mu = 0.1, \sigma = 1$	0.1	2.1	0.1	1

Which asset would you prefer?

Nr 1 as we wish to avoid unnecessary risk



# Variance of Returns: A binomial Model

Pairwise comparison

$S_0 = 1$	50%	50%
$R_0$	$\mu - \sigma$	$\mu + \sigma$

	unlucky (50%)	lucky (50%)	$\mathbb{E}(R_0)$	$\text{Var}(R_0)$
1) $\mu = 0.1, \sigma = 0$	1.1	1.1	0.1	0
3) $\mu = 0.07, \sigma = 0$	1.07	1.07	0.07	0

Which asset would you prefer?

Nr 1 as it yields more profit with the same risk

# Variance of Returns: A binomial Model

Pairwise comparison:

$S_0 = 1$	50%	50%
$R_0$	$\mu - \sigma$	$\mu + \sigma$

	unlucky (50%)	lucky (50%)	$\mathbb{E}(R_0)$	$\text{Var}(R_0)$
1) $\mu = 0.1, \sigma = 0$	1.1	1.1	0.1	0
4) $\mu = 0.2, \sigma = 0.2$	1.0	1.4	0.2	0.04

Which asset would you prefer?

Undecided. 4) has more risk, but rewards by a larger expectation

# Variance of Returns: A binomial Model

Pairwise comparison:

$S_0 = 1$	50%	50%
$R_0$	$\mu - \sigma$	$\mu + \sigma$

	unlucky (50%)	lucky (50%)	$\mathbb{E}(R_0)$	$\text{Var}(R_0)$
2) $\mu = 0.1, \sigma = 1$	0.1	2.1	0.1	1
4) $\mu = 0.2, \sigma = 0.2$	1.0	1.4	0.2	0.04

Which asset would you prefer?

4) has a larger expectation and less risk

# Variance of Returns: A binomial Model

Pairwise comparison:

$S_0 = 1$	50%	50%
$R_0$	$\mu - \sigma$	$\mu + \sigma$

	unlucky (50%)	lucky (50%)	$\mathbb{E}(R_0)$	$\text{Var}(R_0)$
4) $\mu = 0.2, \sigma = 0.2$	1.0	1.4	0.2	0.04
5) $\mu = 0.15, \sigma = 0.1$	1.05	1.25	0.1	0.01

Which asset would you prefer?

Undecided. 5) has less risk, but also  
a lower expectation

## Variance of Returns: A binomial Model

	unlucky (50%)	lucky (50%)	$\mathbb{E}(R_0)$	$\text{Var}(R_0)$
1) $\mu = 0.1, \sigma = 0$	1.1	1.1	0.1	0
2) $\mu = 0.1, \sigma = 1$	0.1	2.1	0.1	1
3) $\mu = 0.07, \sigma = 0$	1.07	1.07	0.07	0
4) $\mu = 0.2, \sigma = 0.2$	1.0	1.4	0.2	0.04
5) $\mu = 0.15, \sigma = 0.1$	1.05	1.25	0.15	0.01

- (1) dominates (2) and (3) and is dominated by no other
- (2) dominates no other, but is dominated by (1), (4) and (5)
- (3) dominates no other and is dominated by (1)
- (4) dominates (2) and is dominated by no other
- (5) dominates (2) and is dominated by no other

If you had to choose one investment, only (1), (4) or (5) are reasonable

# Investment Dominance

Investment 1 (mean  $\mu_1$ , SD  $\sigma_1$ ) **strictly dominates** investment 2 (mean  $\mu_2$ , SD  $\sigma_2$ ), if

$$\mu_1 \geq \mu_2$$

$$\sigma_1 \leq \sigma_2$$

with one of the inequalities being **strict** (i.e.  $\neq$ ).

We write  $(\mu_1, \sigma_1) \succ (\mu_2, \sigma_2)$

At least one of these conditions:

(a)

$$\mu_1 > \mu_2$$

$$\sigma_1 \leq \sigma_2$$

(b)

$$\mu_1 \geq \mu_2$$

$$\sigma_1 < \sigma_2$$

Note: this is NOT Stochastic dominance!

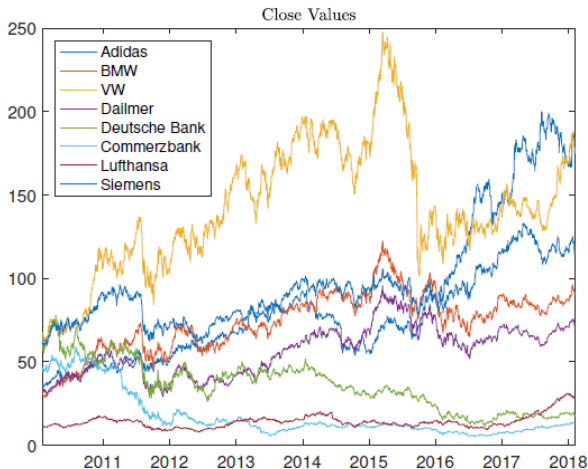
# Efficient Subset

Investments that are not dominated form the *efficient subset*:

The **efficient subset**  $A_{\text{eff}}$  of  $A = \{(\mu_i, \sigma_i), i \in \mathcal{J}\}$  consists of **all elements**  $(\mu_i, \sigma_i), i \in \mathcal{J}$ , which are **not dominated** by any other element of  $A$ .

I.e.  $(\hat{\mu}, \hat{\sigma}) \in A_{\text{eff}}$  iff  $(\hat{\mu}, \hat{\sigma}) \in A$  and for **no**  $i \in \mathcal{J}$  it holds  $(\mu_i, \sigma_i) \succ (\hat{\mu}, \hat{\sigma})$ .

## Example data

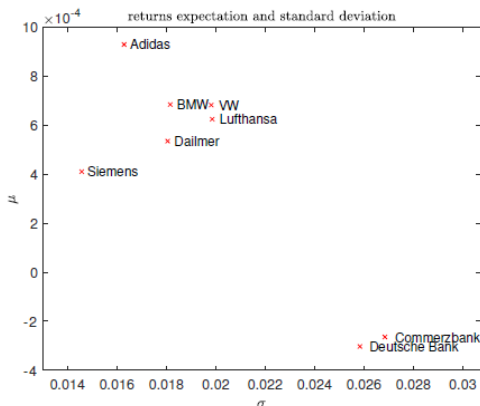




If we can estimate the means and the standard deviations of returns from the data and plot them on a diagram:

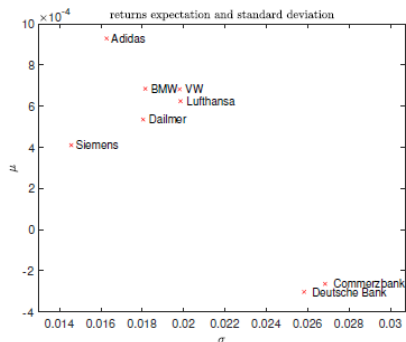
- ▶  $x$  axis: standard deviation
- ▶  $y$  axis: mean or expected return

# Looking at Data



Warning: these values are based on historic prices. There is no guarantee that these are also the future mean and variance!

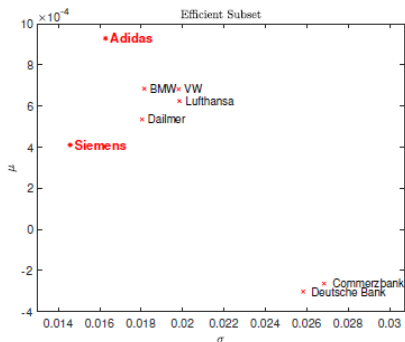
# Looking at Data



## Sample Exam Question:

As an investor, you want to minimise risk and maximise your profit. Based on the given data, which stocks would you invest in, i.e. determine the efficient subset?

# Looking at Data



Eliminating all stocks that are dominated,  
**Adidas** and **Siemens**  
remain as the efficient subset.

# Variance of Returns

Some drawbacks of the variance  $\text{Var}(R) = \mathbb{E}((R - \mathbb{E}(R))^2)$  :

- unexpected large profit contributes same as a loss

Remember:

	25%	25%	50%	mean	variance
Asset 1	£ 750	£1,000	£ 1,175	£ 1,025	30,312.5
Asset 2	£ 800	£1,000	£ 1,500	£ 1,200	95,000

- we cannot distinguish between frequent small losses and a rare huge loss
  - how likely are large losses?
  - how large are likely losses?
- Variance follows historical prices
  - what about events not present in historic prices (e.g. huge bank crash / political changes)

# Semi-Variance of Return

The (*downside*) *semi-variance* of return (*SV*) is defined as:-

$$SV(X) = \sum_{x_i \leq \mu} (x_i - \mu)^2 p_i \text{ if } X \text{ is discrete}$$

$$SV(X) = \int_{-\infty}^{\mu} (x - \mu)^2 f(x) dx \text{ if } X \text{ is continuous}$$

- ▶ It doesn't take into account the variability above the mean ('upside risk')
- ▶ It is not so easy to handle mathematically
- ▶ How does this relate to variance?

# Example

Continuing the first example:

$$\begin{aligned}SV(X) &= \int_{-0.05}^{0.05} 750 \left(0.01 - (x - 0.05)^2\right) (x - 0.05)^2 dx \\&= 750 \int_{-0.05}^{0.05} \left(0.01 (x - 0.05)^2 - (x - 0.05)^4\right) dx \\&= 750 \left[ \frac{0.01}{3} (x - 0.05)^3 - \frac{1}{5} (x - 0.05)^5 \right]_{-0.05}^{0.05} = 0.001\end{aligned}$$

Anything strikes you? Why is this the case?

# Shortfall Probabilities

- ▶ ▶ A shortfall probability measures the probability of returns falling below a certain level - the risk of ruin:

$$\sum_{x < L} p_i \text{ if } X \text{ is discrete}$$

$$\int_{-\infty}^L f(x) dx \text{ if } X \text{ is continuous}$$

- ▶  $L$ : the chosen benchmark level
  - ▶ an absolute level required to meet a payment
  - ▶ return on a benchmark fund
- ▶ denoted  $SP(X)$  or  $SF(\text{benchmark})$



# Example

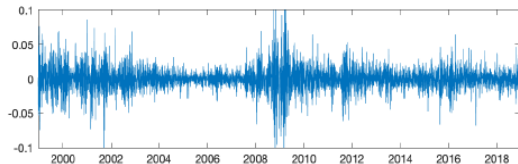
Continuing the example, find the shortfall probability for the stock given that the benchmark return is 0

$$\begin{aligned} SP(X) &= \Pr(X < 0) = \\ &= \int_{-0.05}^0 750 \left( 0.01 - (x - 0.05)^2 \right) dx \\ &= 750 \left[ 0.01x - \frac{(x - 0.05)^3}{3} \right]_{-0.05}^0 = 0.15625 \end{aligned}$$

## Shortfall probability for empirical data

Remember:

Market data:



# Shortfall Probabilities: empirical data

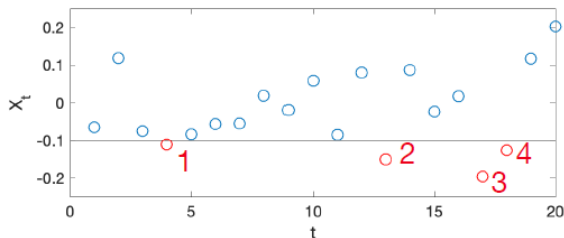
If the benchmark is  $b = -0.1$  :

Count:

Number of large losses vs number of all trading days

$$SF_e(b) = \frac{|\{t : 1 \leq t \leq N, s.t. -X_t > b\}|}{N}$$

Example:



20 samples

4 samples with  $X < -0.1$

$\Rightarrow SF_e(0.1) = 20\%$

# Value at Risk

Value at Risk ( $VaR$ ):

- ▶ statistical measure of the downside risk
- ▶ uses confidence limits to assess the potential losses on a portfolio over a given future time period
- ▶ the largest number  $L$  such that the probability that the loss on the portfolio is greater than  $VaR$ , is  $q$
- ▶ relates to Shortfall Probability but specifies a probability  $q$  and calculates the corresponding shortfall

If  $X$  is discrete:

$$VaR(X; q) = -L \text{ where } L = \{\max x_i : \Pr(X < x_i) \leq q\}$$

If  $X$  is continuous:

$$VaR(X; q) = -L \text{ where } \Pr(X < L) = q$$

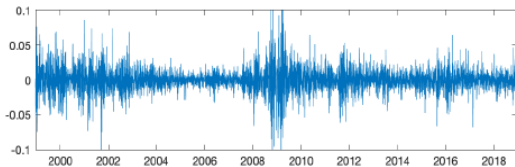
# Value at Risk

- ▶  $VaR$  is the mirror image of  $SP$
- ▶ rather than specify a threshold value  $L$  and measure the probability,  $VaR$  specifies the probability and measures the corresponding threshold value
- ▶  $VaR$  can be calculated from the probability of gains/losses during a period  $T$
- ▶  $VaR$  says: **We are  $100 - q$  certain that we will not lose more than  $\mathcal{L}L$  in time  $T$**
- ▶ Since the nineties  $VaR$  a very popular measure of risk
  - ▶ JPMorgan credited with starting popularising it
  - ▶ Alternative notation  $Var_{\alpha}(X)$  where  $\alpha$  is  $1 - q$

# Value at Risk for empirical data

Again:

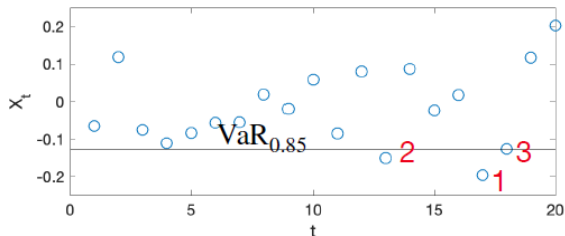
Market data:



# Value at Risk for empirical data

- 1) Sort the values by magnitude
- 2) Consider the smallest  $(1 - \alpha)N_{\text{samples}}$  elements and choose the value of the largest one:

Example:  $\text{VaR}_{0.85}$



20 samples;  
sort by magnitude  
and count the smallest  
 $0.15 \cdot 20 = 3$  entries.

$$\text{VaR}_{0.85} \approx 0.127$$

## Example

Find the  $VaR$  over one year with 95% confidence interval for a portfolio consisting of £100 million invested in the stock used before.

$$\Pr(X < L) = 0.05$$

$$750 \int_{-0.05}^L \left(0.01 - (x - 0.05)^2\right) dx = 0.05$$

$$750 \left(0.01x - \frac{1}{3}(x - 0.05)^3\right)_{-0.05}^L = 0.05$$

$$L = -0.02293$$



## Example

Since  $L$  is a percentage investment return, the 95% value at risk on a £100 million portfolio is  $£100 \text{ million} \times 0.02293 = £2.293 \text{ million}$ .

Interpretation: we are 95% certain that we will not lose more than  $£2.293 \text{ million}$ .

# Expected Shortfall

- ▶ *VaR* asks the question:
  - ▶ How bad things can go?
- ▶ Suppose a bank tells a trader that the one day 99% *VaR* of the trader's portfolio must be kept at less than £10 million.
  - ▶ he constructs a portfolio where 99.1% chance the daily loss is £10 million and 0.9% chance is £500 million: *Unacceptable risk*
- ▶ Expected shortfall asks the question:
  - ▶ If things go bad, what is the expected loss?

# Expected Shortfall

For a shortfall probability  $q$  and corresponding threshold  $L$  such that  $\Pr(X < L) = q$  then expected shortfall is:

$$E[\max(L - X, 0)] = \sum_{x_i \leq L} (L - x_i) p_i \text{ for } X \text{ discrete}$$

$$E[\max(L - X, 0)] = \int_{-\infty}^L (L - x) f(x) dx \text{ for } X \text{ continuous}$$

For the  $(1 - q) \times 100\%$  confidence limit, expected shortfall represents the expected loss in excess of the  $q$ -th lower tail value.

## Example

Find the expected shortfall over one year with 95% confidence interval for a portfolio consisting of £100 million invested in the stock from the initial example.

$$\begin{aligned} & E(-0.02293 - X | X < -0.02293) \\ = & 750 \int_{-0.05}^{-0.02293} (-0.02293 - x) (0.01 - (x - 0.05)^2) dx \\ = & 0.000462 \end{aligned}$$

On a portfolio of £100 million the 95%  $EXSP = £100$  million  $\times 0.000462 = £0.0462$  million.

Interpretation: the expected loss in excess of £2.293 million is £46,200.