

# Return and risk

## Lecture 1

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April 23, 2014



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# Section 1

## Notation

$R$	return (a random variable)
$r_i$	realization of $R$ if $R$ is a discrete variable
$r$	realization of $R$ if $R$ is a continuous variable
$E[R] = \mu_R$	expected value of $R$
$\text{VAR}[R] = \sigma_R^2$	variance of $R$
$\sigma_R$	standard deviation of $R$

## Section 2

### Defining return

# Defining return

## Definition (Return)

Return is the growth rate of the investor's fortune.

# Defining return

Time as discrete variable

$$V_t = V_{t-1} \cdot (1 + r_t)$$

$$r_t = \frac{V_t}{V_{t-1}} - 1$$

Time as a continuous variable

$$V(t) = V(t-1) \cdot \exp(r)$$

$$\exp(r) = \lim_{n \rightarrow \infty} \left(1 + \frac{r}{n}\right)^n$$

$$r = \ln(V(t)) - \ln(V(t-1))$$

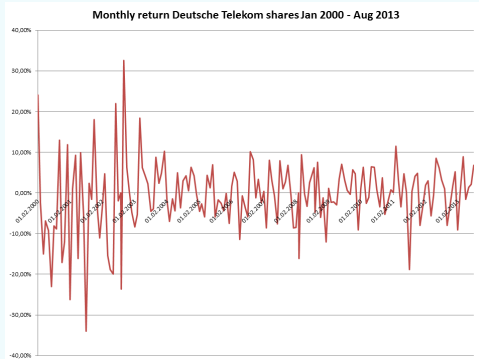
## Section 3

### Past return



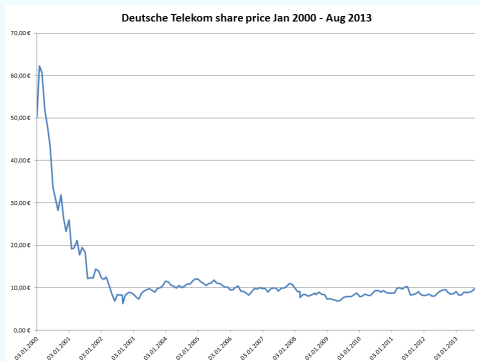
# Past return

Development of returns and share prices over time



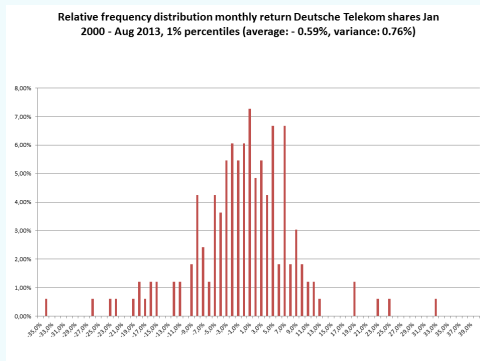
# Past return

Development of returns and share prices over time



# Past return

## Frequency distribution



# Past return

## Measuring past performance

### Measuring past performance

- ▶ Measure by geometric return
- ▶ Average return is misleading

### Example

Sequence of returns: +50%, +30%, -100%

# Past return

## Measuring past performance

### Definition (geometric average)

$$\left( \prod_{t=1}^n (1 + r_t) \right)^{\frac{1}{n}} - 1 = \sqrt[n]{\prod_{t=1}^n (1 + r_t)} - 1$$

In a limited liability economy return cannot be less than -100%.

## Section 4

### Future return

# Future return

## Randomness

### Thesis

The future return  $R$  is a random variable.

# Future return

## Randomness

Randomness	Unpredictability
No objective reason for an event	Determinism. Cause and effect simply to complex to be understood.



# Future return

## Randomness

### Definition (random variable $R$ )

With return as a continuous variable:

$$\begin{aligned} R: \Omega &\rightarrow \{r \in \mathbb{R} \mid r \geq -1\} \\ \omega &\mapsto R(\omega) \end{aligned}$$

With return as a discrete variable:

$$\begin{aligned} R: \Omega &\rightarrow \{r_0, r_1, \dots, r_n\} \\ \omega &\mapsto R(\omega) \end{aligned}$$

The sample space  $\Omega$  is a set of outcomes  $\omega$ .

# Future return

## Randomness

### Examples for discrete sample spaces

Random experiment	Sample space
'weather'	$\Omega = \{\text{sunshine, rain}\}$
'business cycle'	$\Omega = \{\text{expansion, boom, recession, depression}\}$
'stability of currency'	$\Omega = \{\text{currency persists, currency reform}\}$
'play dice'	$\Omega = \{1, 2, 3, 4, 5, 6\}$

# Future return

## Randomness

### Example (Random variable)

$$R(\omega) = \begin{cases} 10\% & \text{if } \omega = \text{'sunshine' } \\ -5\% & \text{if } \omega = \text{'rain' } \end{cases}$$

# Future return

## Expected value

### Definition (expected value of return)

- ▶ with discrete sample space:

$$E[R] = \sum_{\omega \in \Omega} p(\omega) \cdot R(\omega)$$

# Future return

## Expected value

### Definition (expected value of return)

- ▶ with return as a discrete variable:

$$E[R] = \sum_{i=0}^n p(R = r_i) \cdot r_i$$

- ▶ with return as a continuous variable ( $f$  is a probability density function):

$$E[R] = \int_{-\infty}^{+\infty} f(r) \cdot r \, dr$$

# Future return

## Variance

### Thesis

The variance  $\text{VAR}$  of the return  $R$  is a measure of the risk attached to an individual asset.

# Future return

## Variance

### Definition (variance)

- ▶ with return as a discrete variable:

$$\begin{aligned}\text{VAR}(R) &= \text{E}[R] [(R - \mu_R)^2] \\ &= \sum_{i=1}^n p(R = r_i) \cdot (\mu_R - r_i)^2\end{aligned}$$

Remember notation:  $\text{E}(R) = \mu_R$

# Future return

## Variance

### Definition (variance)

- ▶ with continuous sample space ( $f$  is the probability density function):

$$\begin{aligned}\text{VAR}[R] &= \text{E} [(R - \mu_R)^2] \\ &= \int_{-\infty}^{+\infty} f(r) \cdot (r - \mu_R)^2 dr\end{aligned}$$



# Future return

## Probability density

### Definition (Probability density function)

$P$  is a measure for probability. The return  $R$  is supposed to be a **continuous** random variable taking real numbers.

A function  $f : \mathbb{R} \rightarrow [0, \infty[$  is called probability density function of  $R$  if

$$P(r_0 \leq R \leq r_1) = \int_{r_0}^{r_1} f(r) \, dr$$

for all real numbers  $r_0 < r_1$ .

# Future return

## Probability density

In a limited liability economy,  $R$  cannot be less than -100%.

The domain of the probability density function of  $R$  is  $[-1, \infty[$ :

$$f : [-1, \infty[ \rightarrow [0, \infty[$$

It is certain that the random return will take some value between -100% and infinity:

$$\int_{-1}^{\infty} f(r) dr = 1$$

# Future return

## Probability density

### Normal distribution of returns

**Returns** are often assumed to be normally distributed.

Advantage: The normal distribution  $\mathcal{N}(\mu_R, \sigma_R^2)$  is described by just two parameters (namely  $\mu_R$  and  $\sigma_R^2$ ).

# Future return

## Probability density

### Normal distribution of returns

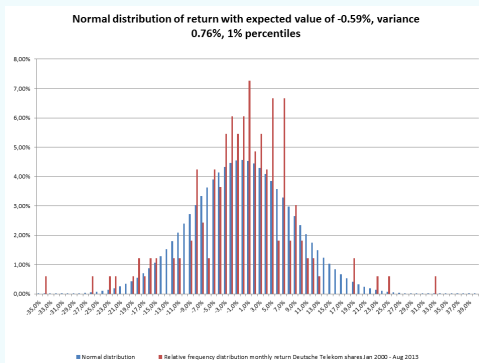
Problem:

- ▶ The domain of a normally distributed variable is  $\mathbb{R}$   
( $f : \mathbb{R} \rightarrow [0, \infty[$ ).
- ▶ But in a limited liability economy,  $R$  cannot be less than -100%, i.e. the domain of the probability density function of  $R$  should be  $[-1, \infty[$ .

# Future return

## Probability density

$\mu_R$  and  $\sigma_R^2$  estimated on the basis of past frequency distributions.



# Future return

## Probability density

### Log-normal distribution of share prices

#### Assumptions:

- ▶ The future return  $R$  is a continuous random variable.
- ▶ The **current** share price  $V_0 = 1$  is a fixed, observed value.
- ▶ The **future** share price  $V_1$  is a random variable.
- ▶  $V_1 = V_0 \cdot \exp(r)$  applies.
- ▶  $R$  is normally distributed.

#### Conclusion:

$V_1$  is log-normally distributed.

# Future return

## Probability density

