

DATASCI 385 Quantitative Finance

Lecture 5: Return and risk

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Suggested reading: Investments Ch 5

Lecture plan

- Risk and risk premiums
- Learning from historical data

Risk

- Any investment involves some **uncertainty** about future holding-period returns, and in many cases that uncertainty is considerable
- Sources of investment risk
 - macroeconomic fluctuations
 - changing fortunes of various industries
 - firm-specific unexpected developments



Holding-period returns

- The *holding-period return*, or HPR (rate of return over the entire time) equals

$$\text{HPR} = \frac{\text{Ending price} - \text{Beginning price} + \text{Cash dividend}}{\text{Beginning price}}$$

$$= \underbrace{\frac{\text{Ending price} - \text{Beginning price}}{\text{Beginning price}}}_{\text{rate of capital gains}} + \underbrace{\frac{\text{Cash dividend}}{\text{Beginning price}}}_{\text{dividend yield}}$$

- Suppose the initial cost of investment is \$100. The end-of-year price is \$110 and cash dividends over the year amount to \$4. Then HPR of a year is

$$\text{HPR} = \frac{110 - 100 + 4}{100} = 0.14$$

Probability distribution of the HPR

- There is considerable uncertainty about share prices one year from now, so you cannot be sure about your eventual HPR

	A	B	C	D	E	F	G	H	I
1									
2									
3	Purchase Price =		\$100			T-Bill Rate =		0.04	
4									
5									
6	State of the		Year-End	Cash		Deviations	Squared		Squared
7	Economy	Probability	Price	Dividends	HPR	from Mean	Deviations	Excess	Deviations
8	Boom	0.25	126.50	4.50	0.3100	0.2124	0.0451	0.2700	0.0451
9	Normal growth	0.45	110.00	4.00	0.1400	0.0424	0.0018	0.1000	0.0018
10	Mild recession	0.25	89.75	3.50	−0.0675	−0.1651	0.0273	−0.1075	0.0273
11	Severe recession	0.05	46.00	2.00	−0.5200	−0.6176	0.3815	−0.5600	0.3815
12	Expected Value (mean)	SUMPRODUCT(B8:B11, E8:E11) =				0.0976			
13	Variance of HPR	SUMPRODUCT(B8:B11, G8:G11) =				0.0380			
14	Standard Deviation of HPR	SQRT(G13) =				0.1949			
15	Risk Premium	SUMPRODUCT(B8:B11, H8:H11) =				0.0576			
16	Standard Deviation of Excess Return	SQRT(SUMPRODUCT(B8:B11, I8:I11)) =				0.1949			

Mean of the HPR

- **Mean return, $E(r)$:** probability-weighted average of the rates of return in each scenario

$$E(r) = \sum_s p(s)r(s)$$

- $p(s)$: the probability of scenario s
- $r(s)$: the HPR in scenario s
- $E(r) = (.25 \times .31) + (.45 \times .14) + [.25 \times (-.0675)] + [.05 \times (-.52)] = .0976$

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Variance of the HPR

- Variance**, σ^2 : the expected value of the *squared* deviation from the mean

$$Var(r) = \sigma^2 = \sum_s p(s)[r(s) - E(r)]^2$$

$$\sigma^2 = .25(.31 - .0976)^2 + .45(.14 - .0976)^2 + .25(-.0675 - .0976)^2 + .05(-.5 - .0976)^2 = .0380$$

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Excess return and risk premium

- **Excess return**: difference between **HPR** and risk-free rate r_f

$$r(s) - r_f$$

- **Excess return** is random
- **Risk-free rate** is **nonrandom** (T-bills rate)
- **Risk premium**: difference between **expected HPR** and risk-free rate r_f

$$E(r) - r_f$$

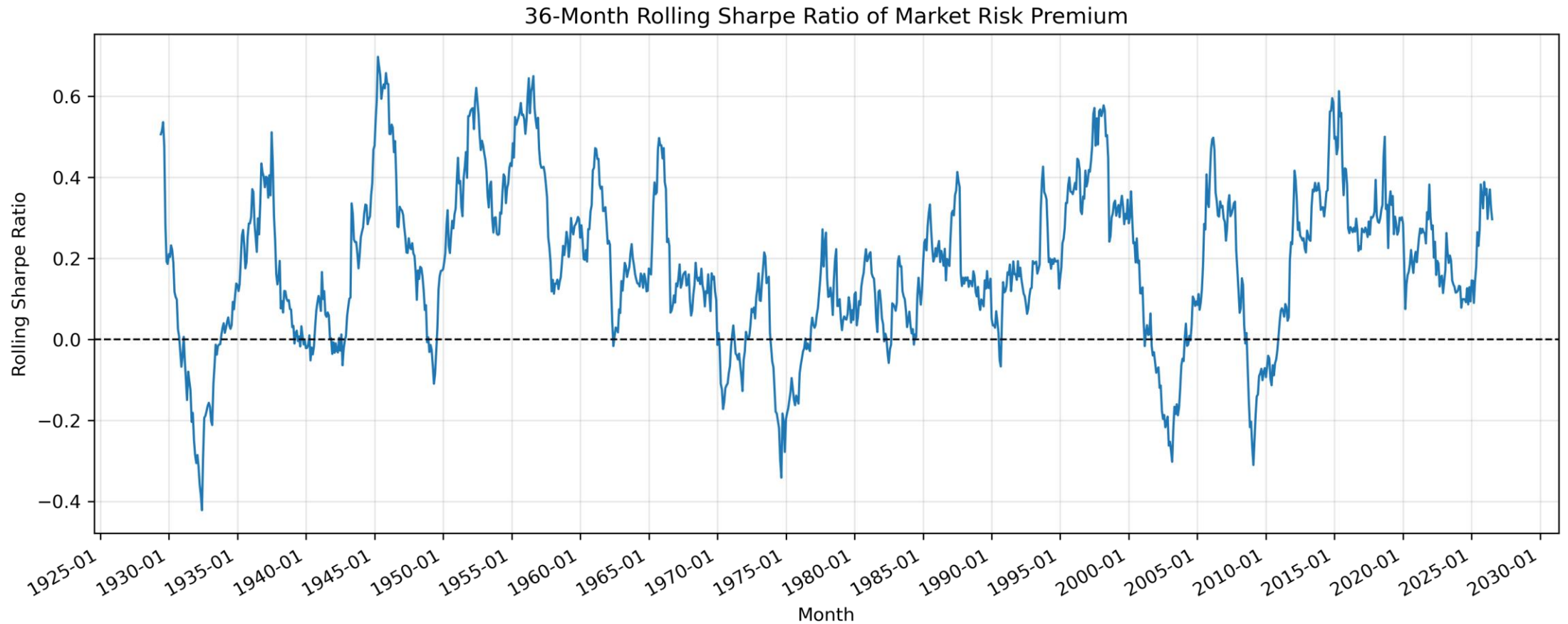
- **Risk premium** is nonrandom

Sharpe ratio

- Higher **risk premium** is always associated with **higher risk** (measured by SD)
- **Sharpe ratio** (reward-to-volatility ratio): risk premium divided by standard deviation

$$\begin{aligned}\text{Sharpe ratio} &= \frac{\text{Risk premium}}{\text{SD of excess return}} \\ &= \frac{E(r) - r_f}{\sigma}\end{aligned}$$

Example: Sharpe ratio of market factor



Lecture plan

- Risk and risk premiums
- Learning from historical data

Expected returns and arithmetic average

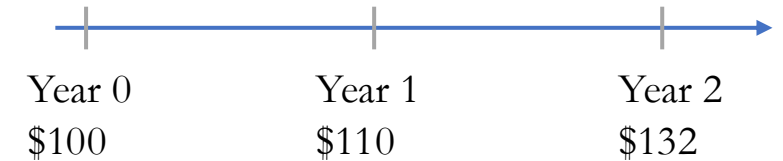
- When we use historical data, we treat each observation as an equally likely “scenario”. If there are n observations,

$$E(r) = \sum_s p(s)r(s) = \frac{1}{n} \sum_{i=1}^n r_i$$

- Arithmetic average of historic rates of return

- Example

- $r_1 = \frac{110}{100} - 1 = 0.1$
- $r_2 = \frac{132}{110} - 1 = 0.2$
- $E(r) = \frac{(0.1+0.2)}{2} = 0.15$



However, $100 \times (1 + 0.15)^2 = 132.25 \neq 132$

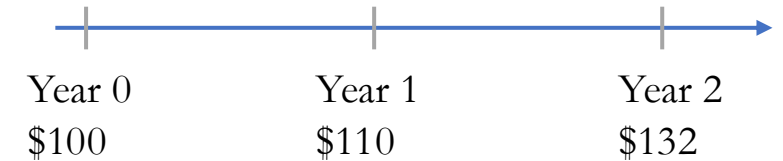
The geometric (time-weighted) average return

- *Geometric* or *compound* rate of return g : the **fixed HPR** that would compound to the same terminal value resulting from the sequence of actual returns in the time series (P_0 : *Initial value*; P_n : *Terminal value*)

$$P_0 \times (1 + r_1) \times (1 + r_2) \times \dots \times (1 + r_n) = P_n$$

$$(1 + g)^n = \frac{P_n}{P_0}$$

- *time-weighted* average return



- Example
 - $(1 + g)^n = 132/100$
 - $g = \left(\frac{132}{100}\right)^{1/2} - 1 = 14.89\%$

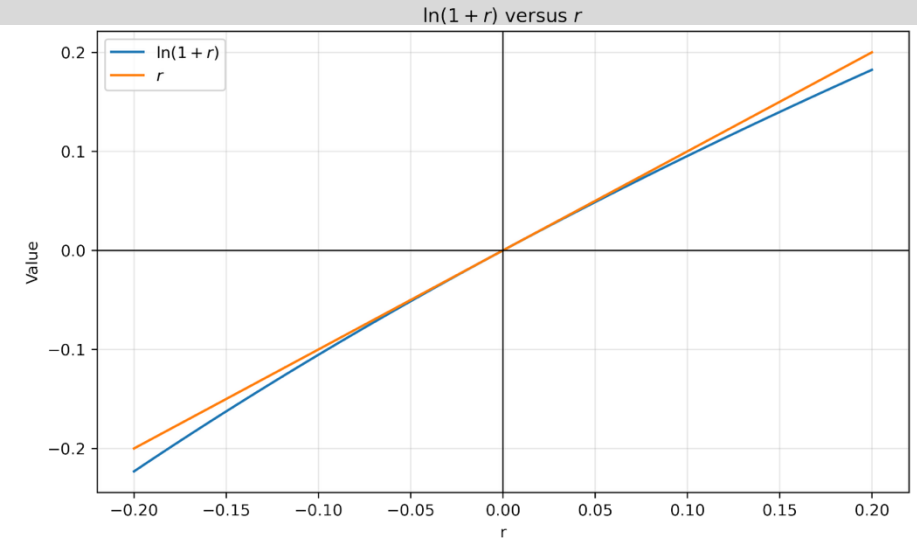
Log return

- When r is small, $\ln(1 + r) \approx r$
- Log return is $r_t = \ln\left(\frac{P_t}{P_{t-1}}\right)$
- Time additivity of r_t ,

$$\ln(P_0) + (r_1 + r_2 + \cdots + r_n) = \ln(P_n)$$

- Let average return be $\bar{r} = (r_1 + r_2 + \cdots + r_n)/n$. Then

$$n\bar{r} = \ln\left(\frac{P_n}{P_0}\right) \text{ or } P_0 e^{\bar{r}} \times e^{\bar{r}} \times \cdots \times e^{\bar{r}} = P_n$$



Estimating variance and standard deviation

- We can estimate the variance of the actual returns from historical data

$$\hat{\sigma}^2 = \frac{1}{n-1} \sum_{s=1}^n (r(s) - \bar{r})^2$$

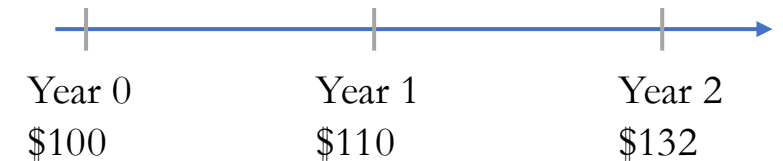
- $\bar{r} = E(r) = \frac{1}{n} \sum_{s=1}^n r(s)$

- The denominator $n - 1$ is to account for the degree of freedom

- Example

- $\hat{\sigma}^2 = \frac{1}{2-1} [(0.2 - 0.15)^2 + (0.1 - 0.15)^2] = 0.005$

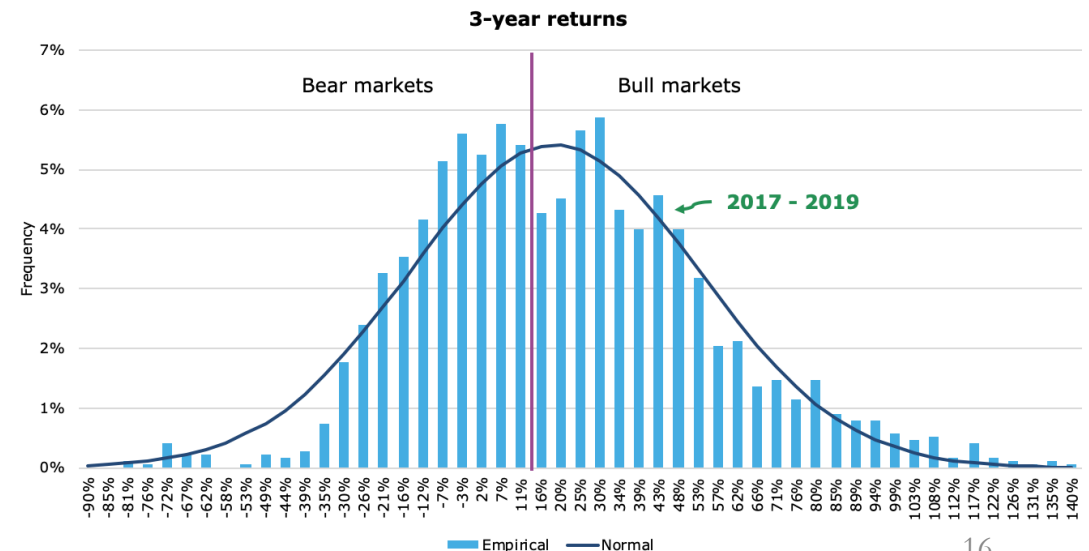
- Standard deviation $\hat{\sigma} = 0.005^{1/2} = 0.071$



Normal approximation of distribution of returns

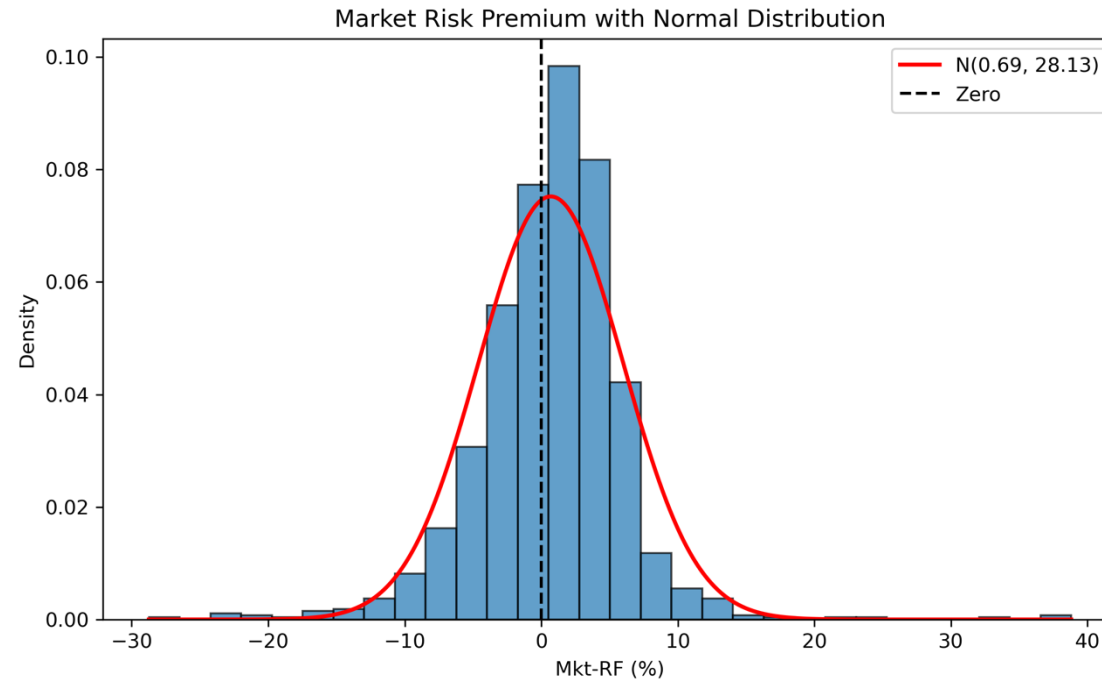
- Normal distribution appears naturally in many applications
 - E.g., the heights and weights of newborns, lifespans of many customers items
- Why does investment management use normal distribution?
 - **Symmetry**: standard deviation is sufficient to capture the risk
 - Scenario analysis is simpler: only **mean** and **variance** are sufficient to estimate scenario probability
 - Easy to model **statistical dependence** of returns across assets: **correlation** is sufficient

- Let us look at the real data...



Deviation from normality

- First deviation: *Asymmetry* in the probability distribution of returns
- Second deviation: Likelihood of *extreme values* on either side of the mean



Measuring asymmetry

- **Skew** measures asymmetry in the probability distribution of returns

$$\text{Skew} = \text{Average} \left[\frac{(r - \bar{r})^3}{\hat{\sigma}^3} \right]$$

- E.g., large negative returns are more likely than large positive returns
- **Negative skew**: extreme *bad* outcomes are **more frequent** than extreme positive ones (skew to the left, fatter left tail, underestimate risk)
- **Positive skew**: opposite case (skew to the right)

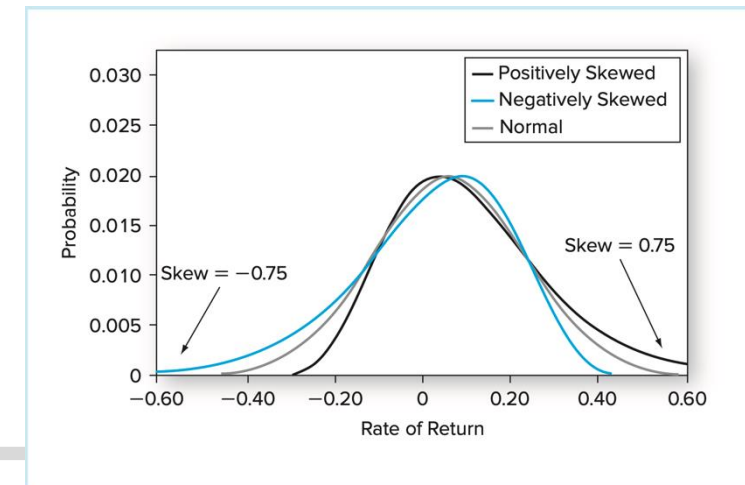


Figure 5.4 Normal and skewed distributions
(mean = 6%, SD = 17%)

Measuring extreme values

- **Kurtosis** measures likelihood of extreme values on either side of the mean

$$\text{Kurtosis} = \text{Average} \left[\frac{(r - \bar{r})^4}{\hat{\sigma}^4} \right] - 3$$

- Deviations are raised to the fourth power so more sensitive to extreme outcomes
- Subtract by **3** because the **kurtosis** for the **normal** distribution is **3**

