

# **Lecture Note:**

# **Quantitative Methods in Finance**

## GENERAL EDUCATION SEMINAR (PEARL)

---

Teruo Nakatsuma

Faculty of Economics, Keio University

# Python

- Python is a high-level programming language.
- Designed by Guido van Rossum
- Released in 1991
- Python is popular.

<https://spectrum.ieee.org/computing/software/the-2018-top-programming-languages>

<https://www.tiobe.com/tiobe-index/>

# Why Python?

- It is free.
- It is slow in execution but highly manageable.
- Python codes are arguably more readable than other languages such as C/C++.
- Numerous packages have been developed for Python.
- Most of them are free and written in faster programming languages such as C/C++.

# How To Obtain Python

- The official Python is downloadable at <https://www.python.org>
- Unfortunately, the plain Python does not include any useful tools for statistics / data science.
- Python distributions for scientific computing
  - **Anaconda**  
<https://www.anaconda.com>
  - **ActivePython**  
<https://www.activestate.com/activepython>
  - **Canopy**  
<https://www.enthought.com/product/canopy>

# Tools For Python Programming

- REPL (Read-Eval-Print-Loop)
  - Terminal-based REPL – **IPython**, **QtConsole**
  - Browser-based REPL – **Jupyter Notebook**  
`https://jupyter-notebook.readthedocs.io/en/latest/`
- An **integrated development environment (IDE)** is an application that consists of integrates an editor, a debugger, a profiler and other tools for developers.
  - **Spyder**  
`https://www.spyder-ide.org/`
  - **PyCharm**  
`https://www.jetbrains.com/pycharm/`

# Basic Packages

- **NumPy** – n-dimensional array and mathematical functions (<https://www.numpy.org>)
- **SciPy** – functions for scientific computing (<https://www.scipy.org>)
- **Matplotlib** – 2D/3D plotting (<https://matplotlib.org>)
- **Pandas** – data structure (<https://pandas.pydata.org>)

# Interest Rate $i$

Suppose  $W(t)$  is the wealth at time  $t$  ( $t \geq 0$ ) and  $r$  is an interest rate.

- Simple interest rate

$$W(t) = (1 + rt)W(0).$$

- One-year compound interest rate

$$W(t) = (1 + r)^t W(0).$$

## Interest Rate ii

- $\frac{1}{M}$ -year compound interest rate

$$W(t) = \left(1 + \frac{r}{M}\right)^{Mt} W(0).$$

Napier's constant is defined as

$$e = \lim_{x \rightarrow \infty} \left(1 + \frac{1}{x}\right)^x.$$



## Interest Rate iii

Then

$$\begin{aligned} & \lim_{M \rightarrow \infty} \left(1 + \frac{r}{M}\right)^M W(0) \\ &= \left\{ \lim_{M \rightarrow \infty} \left(1 + \frac{r}{M}\right)^{\frac{M}{r}} \right\}^r W(0) \\ &= \left\{ \lim_{x \rightarrow \infty} \left(1 + \frac{1}{x}\right)^x \right\}^r W(0) \\ &= e^r W(0), \quad x = \frac{M}{r}. \end{aligned}$$

## Interest Rate iv

Therefore

$$\begin{aligned} W(t) &= \left\{ \lim_{M \rightarrow \infty} \left( 1 + \frac{r}{M} \right)^M \right\}^t W(0) \\ &= \left\{ \lim_{M \rightarrow \infty} \left( 1 + \frac{r}{M} \right)^{\frac{M}{r}} \right\}^{rt} W(0) \\ &= e^{rt} W(0), \end{aligned}$$

which is called the **continuous compound interest rate**.

# Present Value i

## Present value

$$PV(0) = B(t)W(t),$$

where

$$B(t) = \begin{cases} \frac{1}{(1+r)^t}, & \text{(one-year compound),} \\ \frac{1}{(1+\frac{r}{M})^{Mt}}, & \text{(\frac{1}{M}-year compound),} \\ e^{-rt}, & \text{(continuous compound).} \end{cases}$$

## Present Value ii

The **present** value  $PV(0)$  is interpreted as the amount of money you need to put in your bank account now ( $t = 0$ ) to receive  $W(t)$  at time  $t$ . In this context, the interest rate  $r$  is called the **discount rate** and  $B(t)$  is called the **discount factor**.

Conversely,

$$W(t) = \frac{PV(0)}{B(t)},$$

is called the **future value**. The future value is simply regarded as the amount of money you expect to receive at time  $t$ .

# Cash Flow i

A **cash flow** is, as its name suggests, a flow of cash payments. Suppose a project or enterprise (e.g., gold mine, oil well, power plant, factory, shopping mall) will produce cash payments for  $T$  years.

Let  $C(t)$  denote the amount of payment at time  $t$  ( $0 \leq t \leq T$ ). If  $C(t) > 0$ , it is treated as gain or profit (inflow); otherwise, it is loss (outflow). Suppose the number of payments is  $N$  but intervals between them is not necessarily irregular. Let  $t_n$  ( $n = 1, \dots, N$ ) denote the time of the  $n$ -th payment. As a convention, we suppose  $t_0 = 0$  and  $t_N = T$ .

## Cash Flow ii

Then the cash flow is represented as

$$\{C(t_1), \dots, C(t_n), \dots, C(t_N)\}.$$

Furthermore, let  $B(t_n)$  denote the discount factor for  $C(t_n)$ .

**Table 1:** Cash flow, discount factor and present value

Time	$t_1$	$\dots$	$t_n$	$\dots$	$t_N$
Cash Flow	$C(t_1)$	$\dots$	$C(t_n)$	$\dots$	$C(t_N)$
Discount Factor	$B(t_1)$	$\dots$	$B(t_n)$	$\dots$	$B(t_N)$
Present Value	$B(t_1)C(t_1)$	$\dots$	$B(t_n)C(t_n)$	$\dots$	$B(t_N)C(t_N)$

The present value of the cash flow is defined as the sum of all present values, i.e.,

$$\begin{aligned} PV(0) &= B(t_1)C(t_1) + \cdots + B(t_n)C(t_n) + \cdots + B(t_N)C(t_N) \\ &= \sum_{n=1}^N B(t_n)C(t_n). \end{aligned}$$

# Net Present Value i

In practice, we need an initial investment to start the new project that will produce the cash flow  $\{C(t_1), \dots, C(t_N)\}$ . Let  $C(0)$  denote the amount of the initial investment. In case of investment in a new factory, for example,  $C(0)$  represents the amount of fund necessary for the purchase equipments and machineries. Thus  $C(0)$  must be negative.

The **net present value** (NPV) of a cash flow is defined as

$$\begin{aligned} NPV(0) &= C(0) + PV(0) \\ &= C(0) + \sum_{n=1}^N B(t_n) C(t_n) = \sum_{n=0}^N B(t_n) C(t_n). \end{aligned}$$



## Net Present Value ii

Suppose intervals between payments are equal, that is,

$$t_n = \frac{n}{N}T = n\Delta, \quad \Delta = \frac{T}{N}, \quad (n = 0, 1, \dots, N).$$

Then  $B(t_n) = \beta(r)^n$  where

$$\beta(r) = \begin{cases} \frac{1}{(1+r)^\Delta}, & \text{(one-year compound)} \\ \frac{1}{(1+\frac{r}{M})^{M\Delta}}, & (\frac{1}{M}\text{-year compound}) \\ e^{-r\Delta}, & \text{(continuous compound)}. \end{cases}$$

Therefore the NPV is expressed as a polynomial function:

$$NPV(0) = c_0 + c_1x + \cdots + c_nx^n + \cdots + c_Nx^N,$$

where  $x = \beta(r)$  and  $c_n = C(t_n)$  ( $n = 0, 1, \dots, N$ ).

# Internal Rate Of Return i

The **internal rate of return (IRR)** is the discount rate  $r^*$  that makes the NPV of the cash flow equal to zero:

$$0 = c_0 + c_1\beta(r^*) + \cdots + c_n\beta(r^*)^n + \cdots + c_N\beta(r^*)^N.$$

The solution  $r^*$  is obtained by solving the polynomial equation:

$$0 = c_0 + c_1x + \cdots + c_nx^n + \cdots + c_Nx^N,$$

with respect to  $x$ , and transform the solution  $x^*$  with

$$r^* = \begin{cases} \left( \left( \frac{1}{x^*} \right)^{\frac{1}{\Delta}} - 1 \right), & \text{(one-year compound)} \\ M \left\{ \left( \frac{1}{x^*} \right)^{\frac{1}{M\Delta}} - 1 \right\}, & \left( \frac{1}{M} \text{-year compound} \right) \\ -\frac{\log x^*}{\Delta}, & \text{(continuous compound)} \end{cases}$$

## Internal Rate Of Return ii

To simplify mathematical expressions, we suppose  $\Delta = 1$  and  $M = 1$ , i.e., each payment occurs at the end of each year.

### Special cases

**Case 1:** When  $N = 1$ ,

$$0 = c_0 + c_1 x^* \quad \Rightarrow \quad r^* = \frac{c_1 - (-c_0)}{-c_0}.$$

**Case 2:** When  $N \rightarrow \infty$  and  $c_n = c$  for all  $n = 1, 2, 3, 4$ ,

$$0 = c_0 + \sum_{n=1}^{\infty} c(x^*)^n = c_0 + c \frac{x^*}{1 - x^*} \quad \Rightarrow \quad r^* = \frac{c}{-c_0}.$$

## Internal Rate Of Return iii

In general a higher-order polynomial equation could have multiple solutions and many of them are complex numbers. To ensure that the polynomial equation has the unique real-valued solution, the following two conditions must be satisfied:

$$(A1) \ c_0 < 0.$$

$$(A2) \ c_n \geq 0 \ (n = 1, \dots, N) \text{ and } c_n > 0 \text{ for some } n.$$

To ensure that the IRR is positive, we need the extra condition:

$$(A3) \ \sum_{n=0}^N c_n > 0.$$

# Bond

A bond is an instrument of borrowing. A typical bond promises to pay the predetermined amount of cash, the **principal** or the **face value**, to the bond holder on the predetermined future date, the **maturity date**. The length of time until the maturity date is called the **time to maturity**. A **coupon-bearing bond** promises to pay the predetermined amount of cash periodically to the bond holder. This payment is called the **coupon**. A bond without coupon payment is called a **zero-coupon bond**.

## Issuers of bonds

- **Sovereign bond:** a bond issued by a national government
- **Municipal bond:** a bond issued by a local government
- **Corporate bond:** a bond issued by a corporation

# Bond Vs. Bank Loan

As instruments to raise capital, bonds seem similar to bank loans (both are categorized as debts in the balance sheet), though there are notable differences between them:

1. Bonds are often issued at auctions (**primary market**).
2. Bonds can be transferred from one holder to others.
3. Bonds are tradable in the financial market (**secondary market**) until the maturity date.
4. When the bond issuer fails to pay either coupon or principal, it is deemed a **default**. **Credit ratings** are indicators for the default risk of bond issuers, which are assigned by credit rating agencies (e.g., S&P, Moody's, Fitch).

# Cash Flows Of Bonds

## Zero-coupon bond

$t$	0	1	2	...	$T-1$	$T$
$C(t)$	$-V$	0	0	...	0	$F$

## Coupon-bearing bond

$t$	0	1	2	...	$T-1$	$T$
$C(t)$	$-P$	$C$	$C$	...	$C$	$C+F$

$C$  — coupon

$F$  — face value

$T$  — time to maturity

$V$  — price of the zero-coupon bond

$P$  — price of the coupon-bearing bond



# Yield To Maturity Of A Zero-Coupon Bond i

Let  $V(t)$  denote the price of a zero-coupon bond that will mature at time  $t$  ( $0 < t \leq T$ ). The **yield to maturity (YTM)** or simply **yield**, denoted by  $y(t)$ , is defined as

## Zero-coupon bond yield

$$y(t) = \begin{cases} \left( \frac{F}{V(t)} \right)^{\frac{1}{t}} - 1, & \text{(one-year compound)} \\ \frac{1}{t} \log \frac{F}{V(t)}, & \text{(continuous compound)} \end{cases}$$

## Yield To Maturity Of A Zero-Coupon Bond ii

Basically, the yield  $y(t)$  is equivalent to the internal rate of return on the cash flow of the zero-coupon bond.  $y(t)$  as a function of time to maturity is called the **zero yield curve** or simply **yield curve**. Conversely, once the yield curve  $y(t)$  is known, the price of a zero-coupon bond is given by

$$V(t) = \begin{cases} \frac{F}{\{1 + y(t)\}^t}, & \text{(one-year compound)} \\ e^{-y(t)t}F, & \text{(continuous compound)} \end{cases}$$

The price of a zero-coupon bond is equal to the present value of the face value  $F$  discounted by the yield  $y(t)$ .

# Yield To Maturity Of A Coupon-Bearing Bond

The yield to maturity of a coupon-bearing bond is the internal rate of return on its cash flow. In this context,  $y(t)$  must be constant for any  $t$ , though  $y$  is time-varying in reality. Therefore the yield  $y(t)$  is the real-valued solution of the following polynomial equation.

## Coupon-bearing bond yield

$$P(0) = C \sum_{t=1}^{T-1} B(t, y) + (C + F)B(T, y),$$

$$B(t, y) = \begin{cases} \frac{1}{(1 + y)^t}, & \text{(one-year compound)} \\ e^{-yt}, & \text{(continuous compound)} \end{cases}$$

# Relationship Between Bond Price And Bond Yield

## Bond price and bond yield

$$V(y) = B(t, y)F, \quad (t = 1, \dots, T)$$

$$P(y) = C \sum_{t=1}^{T-1} B(t, y) + (C + F)B(T, y).$$

1. Both  $V(t)$  and  $P(y)$  are decreasing functions of  $y$ .
2. The yield will go up when the price goes down.
3. The yield is uniquely determined since (A1) and (A2) are always satisfied.
4. The yield is negative if

$$\sum_{t=0}^T C(t) = TC + F - P < 0.$$

# Price Sensitivity To A Yield Curve Shift i

The yield curve frequently shifts due to business cycles, market sentiments, interventions by central banks and other numerous factors. In this lecture, we concentrate on a parallel shift in the yield curve:  $y(t) + \lambda$  for all  $t$ . We consider the following measurement of sensitivity to a yield curve shift:

$$\text{sensitivity to yield curve shift} = \frac{\lim_{\lambda \rightarrow 0} \frac{P(y+\lambda) - P(y)}{\lambda}}{P(y)},$$

where  $y$  may be time-varying and “ $\lim_{\lambda \rightarrow 0}$ ” means that the shift  $\lambda$  is infinitesimally small.

## Price Sensitivity To A Yield Curve Shift ii

To obtain the exact formula, we need the differential of the bond price with respect to the yield. The differential of  $f(x)$  is

$$\nabla_x f(x) \triangleq \lim_{\epsilon \rightarrow 0} \frac{f(x + \epsilon) - f(x)}{\epsilon}.$$

Using the above notation, we have

$$\begin{aligned} \lim_{\lambda \rightarrow 0} \frac{P(y + \lambda) - P(y)}{\lambda} &= \nabla_y P(y) = C \sum_{t=1}^T \nabla_y B(t, y) + F \nabla_y B(T, y) \\ &= \begin{cases} -C \sum_{t=1}^T \frac{tB(t, y)}{1 + y} - F \frac{TB(T, y)}{1 + y}, & \text{(one-year compound)} \\ -C \sum_{t=1}^T tB(t, y) - FTB(T, y), & \text{(continuous compound)} \end{cases} \end{aligned}$$

## Price Sensitivity To A Yield Curve Shift iii

since

$$\nabla_y B(t, y) = \begin{cases} -\frac{t}{(1+y)^{t+1}}, & \text{(one-year compound)} \\ -te^{-yt}, & \text{(continuous compound)} \end{cases}$$

Hence the price sensitivity to a infinitesimally small shift in the yield curve is given by

$$\frac{\nabla_y P(y)}{P(y)} = \begin{cases} -\frac{D(y)}{1+y}, & \text{(one-year compound)} \\ -D(y), & \text{(continuous compound)} \end{cases}$$
$$D(y) = \frac{C \sum_{t=1}^T tB(t, y) + FTB(T, y)}{P(y)}.$$

# Price Sensitivity To A Yield Curve Shift iv

$D(y)$  is called the **duration**.

## Remarks

1.  $D(y) > 0$ . Thus the price will increase if  $\lambda < 0$ , and vice versa.
2.  $D(y) = t$  for a zero-coupon bond with time to maturity  $t$ .



# Convexity

The **convexity** is defined as

$$C(y) = \frac{\nabla_y^2 P(y)}{P(y)} = \frac{S(y) + (1 + D(y))D(y)}{(1 + y)^2}, \text{ (one-year compound)}$$

$$S(y) = \frac{C \sum_{t=1}^T (t - D(y))^2 B(t, y) + F(T - D(y))^2 B(T, y)}{P(y)}.$$

1. In case of continuous compound, omit  $(1 + y)^2$ .
2.  $S(y)$  is called the **dispersion**.
3.  $C(y) > 0$ . Thus the rate of change in the bond price is a convex function of the shift  $\lambda$ .
4. Suppose there are two bonds with the same duration.  
Then the bond with large convexity will suffer less from the upward shift ( $\lambda > 0$ ) while it will gain more from the downward shift ( $\lambda < 0$ ).

# Yield Curve Estimation i

Since long-term bonds are coupon-bearing ones, we have to estimate the yield curve with price data of coupon-bearing bonds. Suppose  $N$  bonds with various maturities  $T_1, \dots, T_N$  are traded in the market. They must be issued by the same agent. The price of bond  $n$  is  $P_n$  and its cash flow is  $\{C_n(1), \dots, C_n(\bar{T})\}$  where  $\bar{T}$  is the longest maturity in the market. Since bond  $n$  will mature at  $T_n$ ,  $C_n(t) = 0$  for  $t > T_n$ . Then the bond price is given by

$$P_n = \sum_{t=1}^{T_n} B(t)C_n(t) = \sum_{t=1}^{\bar{T}} B(t)C_n(t), \quad (n = 1, \dots, N)$$

where  $B(t)$  is the discount factor.

## Yield Curve Estimation ii

The discount factors  $B(1), \dots, B(\bar{T})$  are obtained as the solution of the following system of equations:

$$\begin{bmatrix} P_1 \\ P_2 \\ \vdots \\ P_N \end{bmatrix} = \begin{bmatrix} C_1(1) & C_1(2) & \cdots & C_1(\bar{T}) \\ C_2(1) & C_2(2) & \cdots & C_2(\bar{T}) \\ \vdots & & \ddots & \vdots \\ C_N(1) & C_N(2) & \cdots & C_N(\bar{T}) \end{bmatrix} \begin{bmatrix} B(1) \\ B(2) \\ \vdots \\ B(\bar{T}) \end{bmatrix}.$$

Finally, the yield curve  $y(t)$  is given by

$$y(t) = \begin{cases} \left( \frac{1}{B(t)} \right)^{\frac{1}{t}} - 1, & \text{(one-year compound)} \\ \frac{1}{t} \log \frac{1}{B(t)}, & \text{(continuous compound)} \end{cases}$$

# Defaultable Bond And Credit Spread i

Suppose the probability of default is  $\pi$  and the recovery rate is  $\rho$  ( $0 \leq \pi, \rho < 1$ ). Then the payment of a defaultable zero-coupon bond is

$$\begin{cases} \rho F, & \text{(the issuer defaults with } \pi) \\ F, & \text{(the issuer does not default with } 1 - \pi) \end{cases}$$

The expected payment is

$$(1 - \pi)F + \pi\rho F = (1 - (1 - \rho)\pi)F.$$

## Defaultable Bond And Credit Spread ii

The “fair price” of the zero-coupon bond is defined as

$$\tilde{V} = \begin{cases} \frac{(1 - (1 - \rho)\pi)F}{(1 + \tilde{y})^T}, & \text{(one-year compound)} \\ e^{-\tilde{y}T}(1 - (1 - \rho)\pi)F, & \text{(continuous compound)} \end{cases}$$

The yield on the defaultable bond is given by

$$\tilde{y} = \begin{cases} \frac{1 + y}{(1 - (1 - \rho)\pi)^{\frac{1}{T}}} - 1, & \text{(one-year compound)} \\ y - \frac{1}{T} \log(1 - (1 - \rho)\pi), & \text{(continuous compound)} \end{cases}$$

where  $y$  is the yield on riskless bond with the same maturity.

## Defaultable Bond And Credit Spread iii

Since  $1 - (1 - \rho)\pi < 1$ ,  $\tilde{y} - y$  is positive. The difference between  $\tilde{y}$  and  $y$  is called the **credit spread**.

1. The credit spread is increased when the probability of default is increased.
2. The credit spread is increased when the recovery rate is decreased.

# Portfolio Analysis

Portfolio = a collection of assets that the investor holds

**Types of assets:** stocks, bonds, currencies, real estate

**Questions to be answered**

- How much we must invest our money in each asset?
- What is the best way to form a suitable portfolio?

# Return On A Portfolio Of Assets

The return on a portfolio of asset is a weighted average of the return on the individual assets.

## Return of a portfolio

$$R_P = \sum_{n=1}^N w_n R_n.$$

$R_P$ : the return on the portfolio

$R_n$ : the return on the  $n$ -th asset

$w_n$ : the fraction of the funds invested in the  $n$ -th asset



# Measure Of Average Outcome

## Expected return

$$\begin{aligned}\mu_P &= E(R_P) = E\left(\sum_{n=1}^N w_n R_n\right) \\ &= \sum_{n=1}^N w_n E(R_n) = \sum_{n=1}^N w_n \mu_n.\end{aligned}$$

The expected return on a portfolio is the weighted average of the expected return on the individual assets.

# Measure Of Dispersion

## Variance

$$\begin{aligned}\sigma_P^2 &= E[(R_P - \mu_P)^2] = E \left[ \left\{ \sum_{n=1}^N w_n (R_n - \mu_n) \right\}^2 \right] \\&= \sum_{n=1}^N w_n^2 E[(R_n - \mu_n)^2] \\&\quad + \sum_{n=1}^N \sum_{m \neq n}^N w_n w_m E[(R_n - \mu_n)(R_m - \mu_m)] \\&= \sum_{n=1}^N w_n^2 \sigma_n^2 + \sum_{n=1}^N \sum_{m \neq n}^N w_n w_m \sigma_{nm}.\end{aligned}$$

## Example: Portfolio Of Two Assets

The expected return on a portfolio of two assets is

$$\mu_P = w_1\mu_1 + w_2\mu_2,$$

and the variance is

$$\begin{aligned}\sigma_P^2 &= w_1^2\sigma_1^2 + w_2^2\sigma_2^2 + 2w_1w_2\sigma_{12} \\ &= w_1^2\sigma_1^2 + w_2^2\sigma_2^2 + 2w_1w_2\rho_{12}\sigma_1\sigma_2,\end{aligned}$$

where  $\sigma_{12} = \rho_{12}\sigma_1\sigma_2$  and  $\rho_{12}$  is the correlation coefficient between  $R_1$  and  $R_2$ .

## Example: Portfolio Of Three Assets

The expected return on a portfolio of three assets is

$$\mu_P = w_1\mu_1 + w_2\mu_2 + w_3\mu_3,$$

and the variance is

$$\begin{aligned}\sigma_P^2 &= w_1^2\sigma_1^2 + w_2^2\sigma_2^2 + w_3^2\sigma_3^2 \\ &\quad + 2w_1w_2\sigma_{12} + 2w_1w_3\sigma_{13} + 2w_2w_3\sigma_{23} \\ &= w_1^2\sigma_1^2 + w_2^2\sigma_2^2 + w_3^2\sigma_3^2 \\ &\quad + 2w_1w_2\rho_{12}\sigma_1\sigma_2 + 2w_1w_3\rho_{13}\sigma_1\sigma_3 \\ &\quad + 2w_2w_3\rho_{23}\sigma_2\sigma_3,\end{aligned}$$

# Limit Of Diversification

If  $w_n = \frac{1}{N}$  for all assets, the variance of the portfolio is

$$\begin{aligned}\sigma_P^2 &= \sum_{n=1}^N \frac{\sigma_n^2}{N^2} + \sum_{n=1}^N \sum_{m \neq n}^N \frac{\sigma_{nm}}{N^2} \\&= \underbrace{\frac{1}{N} \sum_{n=1}^N \frac{\sigma_n^2}{N}}_{\bar{\sigma}_n^2} + \frac{N-1}{N} \underbrace{\sum_{n=1}^N \sum_{m \neq n}^N \frac{\sigma_{nm}}{N(N-1)}}_{\bar{\sigma}_{nm}} \\&= \frac{1}{N} (\bar{\sigma}_n^2 - \bar{\sigma}_{nm}) + \bar{\sigma}_{nm}.\end{aligned}$$

Thus  $\sigma_P^2$  will converge to  $\bar{\sigma}_{nm} \neq 0$  as  $N$  goes to infinity. Thus even well-diversified portfolios are not necessarily riskless.

# Basic Idea Of Portfolio Selection

- **Expected return as the performance measure**

The investors love a higher return from their investment.

- **Variance as the risk measure**

The investors want to avoid a higher variation in the value of their funds.

But we cannot have both.

We need to find a “sweet spot” in the trade-off between risk and return.

# Minimum Variance Portfolio Problem i

$$\begin{aligned} \min_{w_1, \dots, w_N} \quad & \sigma_P^2 = \sum_{n=1}^N w_n^2 \sigma_n^2 + \sum_{n=1}^N \sum_{m \neq n}^N w_n w_m \sigma_{nm}, \\ \text{subject to} \quad & \text{(a) } \sum_{n=1}^N w_n \mu_n = \mu_P, \\ & \text{(b) } \sum_{n=1}^N w_n = 1, \\ & \text{(c) } w_n \geq 0, \quad (n = 1, \dots, N). \end{aligned} \tag{1}$$

## Minimum Variance Portfolio Problem ii

- The constraint (a) sets the target return of the portfolio.
- The constraint (b) is required due to the definition of  $w_1, \dots, w_n$ .
- The constraint (c) prohibits short selling.

The solution of the above problem (1) gives us the trade-off relationship between the risk  $\sigma_P$  and the return  $\mu_P$ .



# Matrix Form

$$\begin{aligned} \min_w \quad & \sigma_P^2 = \mathbf{w}^\top \mathbf{\Sigma} \mathbf{w}, \\ \text{subject to} \quad & \text{(a) } \mathbf{w}^\top \boldsymbol{\mu} = \mu_P, \\ & \text{(b) } \mathbf{w}^\top \boldsymbol{\iota} = 1, \\ & \text{(c) } \mathbf{w} \geq \mathbf{0}, \end{aligned} \tag{2}$$

where

$$\mathbf{w} = \begin{bmatrix} w_1 \\ \vdots \\ w_N \end{bmatrix}, \quad \boldsymbol{\mu} = \begin{bmatrix} \mu_1 \\ \vdots \\ \mu_N \end{bmatrix}, \quad \mathbf{\Sigma} = \begin{bmatrix} \sigma_1^2 & \cdots & \sigma_{1N} \\ \vdots & \ddots & \vdots \\ \sigma_{N1} & \cdots & \sigma_N^2 \end{bmatrix},$$

and  $\boldsymbol{\iota}$  is the  $\mathbf{N} \times \mathbf{1}$  vector whose elements are equal to 1.

# Optimal Portfolio Selection Procedure i

- For each given  $\mu_P$ , the minimum variance  $\sigma_P^2$  is obtained by solving the minimum variance portfolio problem (2). The graph of  $(\sigma_P, \mu_P)$  is called the **minimum variance frontier**.
- In particular, the upper half of the minimum variance frontier is called the **efficient frontier** because the investor prefers a portfolio with a higher rate of return when the risk is the same.

## Optimal Portfolio Selection Procedure ii

The efficient frontier shows us the trade-off between risk and return. The investor must choose the optimal portfolio among those on the efficient frontier.

**Step 1.** Construct the efficient frontier.

**Step 2.** Set the tolerable level of risk or the target expected return.

**Step 3.** Pick the corresponding portfolio on the efficient frontier.

# Solution

If we drop the constraint (c) which prohibits short selling, we can derive the solution of the minimum variance portfolio problem in a closed form:

$$w = \frac{C\mu_P - A}{D}\Sigma^{-1}\mu + \frac{B - A\mu_P}{D}\Sigma^{-1}\iota$$

$$A = \mu^\top \Sigma^{-1} \iota, \quad B = \mu^\top \Sigma^{-1} \mu,$$

$$C = \iota^\top \Sigma^{-1} \iota, \quad D = BC - A^2.$$

The relationship between  $\mu_P$  and  $\sigma_P$  is given by

$$\sigma_P = \sqrt{\frac{C\mu_P^2 - 2A\mu_P + B}{D}} = \sqrt{\frac{C}{D} \left( \mu_P - \frac{A}{C} \right)^2 + \frac{1}{C}}.$$

This is the **minimum variance frontier**.

# Remarks On Minimum Variance Portfolio Selection

1. The minimum variance portfolio problem without short selling (2) has no closed-form solution.
2. Since the minimum variance portfolio problem (2) is a type of quadratic programming problem, it can be solved by a quadratic programming solver.
3. We may incorporate additional conditions (trading fees, taxes, upper and lower bounds of weights, etc.) into the minimum variance portfolio problem (2).

## When $\mu$ And $\Sigma$ Are Unknown i

So far we assume we know  $\mu$  and  $\Sigma$  in (2). In reality, however, we need to estimate them with data. Let  $r_{nt}$  denote realized return on asset  $n$  at period  $t$  ( $t = 1, \dots, T$ ). The sample mean  $\bar{r}_n$  and the sample covariance  $s_{nm}$  (or the sample variance  $s_n^2$  when  $n = m$ ) are defined as

$$\begin{aligned}\bar{r}_n &= \frac{1}{T} \sum_{t=1}^T r_{nt}, \\ s_{nm} &= \frac{1}{T} \sum_{t=1}^T (r_{nt} - \bar{r}_n)(r_{mt} - \bar{r}_m), \\ &\quad (n, m = 1, \dots, N).\end{aligned}$$

## When $\mu$ And $\Sigma$ Are Unknown ii

A straightforward way is to replace  $\mu$  and  $\Sigma$  in (2) with

$$\bar{\mathbf{r}} = \begin{bmatrix} \bar{r}_1 \\ \vdots \\ \bar{r}_N \end{bmatrix}, \quad \mathbf{S} = \begin{bmatrix} s_1^2 & \cdots & s_{1N} \\ \vdots & \ddots & \vdots \\ s_{N1} & \cdots & s_N^2 \end{bmatrix},$$

respectively, but a more elegant method is known in the literature.

# Alternative Expression Of The Portfolio Variance

$$\begin{aligned}w^T S w &= \sum_{n=1}^N \sum_{m=1}^N w_n w_m S_{nm} \\&= \sum_{n=1}^N \sum_{m=1}^N w_n w_m \left\{ \frac{1}{T} \sum_{t=1}^T (r_{nt} - \bar{r}_n)(r_{mt} - \bar{r}_m) \right\} \\&= \frac{1}{T} \sum_{t=1}^T \sum_{n=1}^N w_n (r_{nt} - \bar{r}_n) \sum_{m=1}^N w_m (r_{mt} - \bar{r}_m) \\&= \frac{1}{T} \sum_{t=1}^T \left\{ \sum_{n=1}^N w_n (r_{nt} - \bar{r}_n) \right\}^2 = \frac{1}{T} \sum_{t=1}^T (r_{Pt} - \bar{r}_P)^2, \\r_{Pt} &= \sum_{n=1}^N w_n r_{nt}, \quad \bar{r}_P = \sum_{n=1}^N w_n \bar{r}_n = \frac{1}{T} \sum_{t=1}^T r_{Pt},\end{aligned}$$



# Remarks

- $r_{Pt}$  is a *realized return* of the portfolio with allocation weights  $\mathbf{w}$  at period  $t$
- $\bar{r}_P$  is the average of realized returns of the portfolio with allocation weights  $\mathbf{w}$  from period **1** to period  $T$ .
- $\frac{1}{T} \sum_{t=1}^T (r_{Pt} - \bar{r}_P)^2$  is the sample variance of realized returns of the portfolio with allocation weights  $\mathbf{w}$  from period **1** to period  $T$ .
- Using  $\mathbf{w}^T \mathbf{S} \mathbf{w}$  implies that we use the realized variance of a portfolio in a hypothetical situation; What if we invest our money on a portfolio with allocation weights  $\mathbf{w}$  from period **1** to period  $T$ ?

# Alternative Form Of The Minimization Problem

Define  $\mathbf{v}_t = \mathbf{r}_{Pt} - \bar{\mathbf{r}}_P$ . Then the minimum variance problem with unknown  $\boldsymbol{\mu}$  and  $\boldsymbol{\Sigma}$  is given by

## Minimum variance portfolio problem

$$\begin{aligned} \min_{\mathbf{w}, \mathbf{v}} \quad & \widehat{\text{Var}}[R_P] = \frac{1}{T} \mathbf{v}^\top \mathbf{v}, \\ \text{s.t.} \quad & D\mathbf{w} = \mathbf{v}, \quad \mathbf{w}^\top \bar{\mathbf{r}} = \mu_P, \quad \mathbf{w}^\top \boldsymbol{\ell} = 1, \\ & \mathbf{w}_1 \geq 0, \dots, \mathbf{w}_N \geq 0, \end{aligned} \tag{3}$$

where

$$\mathbf{v} = \begin{bmatrix} \mathbf{v}_1 \\ \vdots \\ \mathbf{v}_T \end{bmatrix}, \quad D = \begin{bmatrix} r_{11} - \bar{r}_1 & \cdots & r_{N1} - \bar{r}_N \\ \vdots & \ddots & \vdots \\ r_{1T} - \bar{r}_1 & \cdots & r_{NT} - \bar{r}_N \end{bmatrix}.$$

# Alternative Risk Criterion

Other than the variance  $E[(R_P - \mu_P)^2]$ , many alternative risk criteria have been proposed in the literature.

In this lecture, we study the following three alternatives:

- **Mean absolute deviation:**  $E[|R_P - \mu_P|]$
- **Semivariance:**  $E[(R_P - \mu_P)^2 | R_P \leq \mu_P]$
- **Expected shortfall:**  
 $E[-R_P | R_P \leq \text{VaR}_\alpha] \text{ (Pr}\{R_P \leq \text{VaR}_\alpha\} = \alpha)$

# Mean Absolute Deviation Optimization

The sample mean absolute deviation is

$$\varrho^{AD}(\mathbf{w}) = \frac{1}{T} \sum_{t=1}^T |r_{Pt} - \bar{r}_P|. \quad (4)$$

Define  $\mathbf{v}_t = \mathbf{r}_{Pt} - \bar{\mathbf{r}}_P$  and use the same notations as in (3).

## Minimum mean absolute deviation portfolio problem

$$\begin{aligned} \min_{\mathbf{w}, \mathbf{v}} \quad & \varrho^{AD}(\mathbf{w}) = \frac{1}{T} \sum_{t=1}^T |\mathbf{v}_t|, \\ \text{subject to} \quad & D\mathbf{w} = \mathbf{v}, \\ & \mathbf{w}^\top \bar{\mathbf{r}} = \mu_P, \quad \mathbf{w}^\top \boldsymbol{\iota} = 1, \\ & \mathbf{w}_1 \geq 0, \dots, \mathbf{w}_N \geq 0, \end{aligned} \quad (5)$$

# Semivariance Optimization

The sample semivariance is

$$\varrho^{SV}(w) = \frac{1}{T} \sum_{t=1}^T \{[r_{Pt} - \bar{r}_P]^{-}\}^2, \quad (6)$$

where  $[r_{Pt} - \bar{r}_P]^{-} = \max\{-(r_{Pt} - \bar{r}_P), 0\}$  is call the **negative part**. Defining  $v_t = [r_{Pt} - \bar{r}_P]^{-}$ , we have

## Minimum semivariance portfolio problem

$$\begin{aligned} \min_{w, v} \quad & \varrho^{SV}(w) = \frac{1}{T} \sum_{t=1}^T v_t^2, \\ \text{subject to} \quad & w^T \bar{r} = \mu_P, \quad w^T \mathbf{1} = 1, \\ & w_1 \geq 0, \dots, w_N \geq 0, \quad v_1 \geq 0, \dots, v_T \geq 0, \\ & r_{P1} - \bar{r}_P + v_1 \geq 0, \dots, r_{PT} - \bar{r}_P + v_T \geq 0, \end{aligned} \quad (7)$$

# Value-At-Risk And Expected Shortfall

The value-at-risk (VaR) of a portfolio is defined as

**VaR**

$$\Pr\{R_P \leq \text{VaR}_\alpha\} = \alpha. \quad (8)$$

The expected shortfall (ES) is defined as

**Expected shortfall**

$$\text{ES}_\alpha = \mathbb{E}[-R_P | R_P \leq \text{VaR}_\alpha] \quad (9)$$

ES is interpreted as the conditional expected loss under a severe market condition in which our portfolio suffers from a rare but huge loss ( $R_P \leq \text{VaR}_\alpha$ ).

# Coherent Risk Measure i

## Coherent risk measure [Artzner et al. (1999)]

Suppose  $\mathcal{X}$  is a set of random variables. We regard each  $X \in \mathcal{X}$  as the return (or value) of a portfolio and let  $\varrho(X)$  denote a risk measure of  $X$ .  $\varrho(\cdot)$  is said to be coherent if it satisfies the following conditions:

**Monotonicity:** For any  $X, Y \in \mathcal{X}$ ,

$$\Pr\{X \leq Y\} = 1 \text{ implies } \varrho(X) \geq \varrho(Y).$$

**Cash invariance:** For  $M \in \mathcal{X}$  such that  $\Pr\{M = R\} = 1$ ,

$$\varrho(X + M) = \varrho(X) - R.$$

**Sub-additivity:** For any  $X, Y \in \mathcal{X}$ ,  $\varrho(X + Y) \leq \varrho(X) + \varrho(Y)$ .

**Positive homogeneity:** For any  $\lambda \geq 0$ ,  $\varrho(\lambda X) = \lambda \varrho(X)$ .

## Interpretations of conditions

**Monotonicity:** A surely profitable portfolio should be less risky.

**Cash invariance:** Adding the riskless asset should reduce the risk.

**Sub-additivity:** Diversification should not make the portfolio riskier.

**Positive homogeneity:** The risk should be proportional to the position.

**Remark:** ES is coherent but VaR is not.



# Expected Shortfall Optimization

If the probability density function of  $R_P$ , say  $p(R_P)$ , is known, the ES is given by

$$ES_\alpha = \frac{1}{\alpha} \int_{-\infty}^{\infty} [R_P - VaR_\alpha]^- p(R_P) dR_P - VaR_\alpha. \quad (10)$$

The portfolio selection problem with the ES is formulated as

## Minimum expected shortfall portfolio problem

$$\begin{aligned} \min_w \quad & \frac{1}{\alpha} \int_{-\infty}^{\infty} [R_P - VaR_\alpha]^- p(R_P) dR_P - VaR_\alpha, \\ \text{subject to} \quad & w^T \bar{r} = \mu_P, \quad w^T \iota = 1, \quad w_1 \geq 0, \dots, w_N \geq 0. \end{aligned} \quad (11)$$

Unfortunately, (11) is a non-linear optimization problem which is difficult to solve unless we introduce additional assumptions such as the normality of asset returns.

# Approximation Method

To circumvent this obstacle, Rockafellar and Uryasev (2000) proposed an approximation method. If  $T$  is large enough, the ES (10) is approximately equivalent to

$$\varrho^{ES}(w, c) = \frac{1}{\alpha T} \sum_{t=1}^T [r_{Pt} - c]^- - c. \quad (12)$$

Thus, by defining  $v_t = [r_{Pt} - c]^-$ , we have

## Minimum expected shortfall portfolio problem

$$\begin{aligned} \min_{w, v, c} \quad & \varrho^{ES}(w, c) = \frac{1}{\alpha T} \sum_{t=1}^T v_t - c, \\ \text{subject to} \quad & w^T \bar{r} = \mu_P, \quad w^T \iota = 1, \\ & w_1 \geq 0, \dots, w_N \geq 0, \quad v_1 \geq 0, \dots, v_T \geq 0, \\ & r_{P1} - c + v_1 \geq 0, \dots, r_{PT} - c + v_T \geq 0. \end{aligned} \quad (13)$$

# Risk Parity Approach

A **risk parity approach** of portfolio management focuses on balancing “risk allocation” among assets, instead of balancing the target return and the risk of the portfolio.

1. The mean-variance approach tends to produce an extremely skewed portfolio.
2. It is hard to obtain a reliable estimate of the expected return on any asset.
3. Many fund managers feel increasingly uncomfortable with the traditional asset allocation techniques since they were useless during such financial turmoil as the Global Financial Crisis.

# Equal-weight ( $1/N$ ) Portfolio

One trivial example of such portfolios that equalize the impact of each asset onto the total risk is

**$1/N$  portfolio**

$$w_n^{1/N} = \frac{1}{N}, \quad (n = 1, \dots, N). \quad (14)$$

This is often referred to as the  **$1/N$  portfolio**.

# Global Minimum Variance Portfolio

The **global minimum variance portfolio** is the solution of the following optimization problem:

$$\begin{aligned} \min_w \quad & w^\top \Sigma w \\ \text{subject to} \quad & w^\top \iota = 1, \end{aligned} \tag{15}$$

that is,

**Global minimum variance portfolio**

$$w^{MV} = \frac{1}{\iota^\top \Sigma^{-1} \iota} \Sigma^{-1} \iota. \tag{16}$$

Note that (16) does not depend on the expected return vector  $\mu$ . Thus we do not need to estimate  $\mu$ .

# Properties i

1. Suppose the variance of the return on the  $n$ -th asset is  $\sigma_n^2$  ( $n = 1, \dots, N$ ). If there is no correlation among the asset returns, the allocation weight in the global minimum variance portfolio is given by

$$w_n^{MV} = \frac{\sigma_n^{-2}}{\sum_{n=1}^N \sigma_n^{-2}}, \quad (n = 1, \dots, N).$$

2. Suppose  $\sigma_1^2 = \dots = \sigma_N^2$  and the correlation coefficient between any pair of assets is constant. Then the global minimum variance portfolio is equivalent to the  $\mathbf{1}/N$  portfolio.

## Properties ii

3. Consider a **Lagrangian** for `pyfin.alt.global.minvar`:

$$\mathcal{L} = \mathbf{w}^\top \Sigma \mathbf{w} + \lambda(1 - \mathbf{w}^\top \boldsymbol{\iota}),$$

where  $\lambda$  is the **Lagrange multiplier**. The first-order condition is

$$\nabla_{\mathbf{w}} \mathcal{L} = \Sigma \mathbf{w} - \lambda \boldsymbol{\iota} = \mathbf{0} \quad \Rightarrow \quad \Sigma \mathbf{w} = \lambda \boldsymbol{\iota}, \quad (17)$$

Note that the first derivative of  $\sigma(\mathbf{w}) = \sqrt{\mathbf{w}^\top \Sigma \mathbf{w}}$  is given by

$$\nabla \sigma(\mathbf{w}) = \begin{bmatrix} \nabla_{w_1} \sigma(\mathbf{w}) \\ \vdots \\ \nabla_{w_N} \sigma(\mathbf{w}) \end{bmatrix} = \begin{bmatrix} \nabla_1 \sigma(\mathbf{w}) \\ \vdots \\ \nabla_N \sigma(\mathbf{w}) \end{bmatrix} = \frac{1}{\sigma(\mathbf{w})} \Sigma \mathbf{w}.$$

Therefore the first-order condition (17) implies

$$\nabla_1 \sigma(\mathbf{w}) = \dots = \nabla_N \sigma(\mathbf{w}).$$

$\nabla_n \sigma(\mathbf{w})$  ( $n = 1, \dots, N$ ) is called the **MRC (Marginal Risk Contribution)**.



# Risk Decomposition

The standard deviation  $\sigma(\mathbf{w})$  is decomposed as

$$\begin{aligned}\sigma(\mathbf{w}) &= \frac{1}{\sigma(\mathbf{w})} \mathbf{w}^\top \Sigma \mathbf{w} = \mathbf{w}^\top \left( \frac{1}{\sigma(\mathbf{w})} \Sigma \mathbf{w} \right) \\ &= \mathbf{w}^\top \nabla \sigma(\mathbf{w}) \\ &= \mathbf{w}_1 \nabla_1 \sigma(\mathbf{w}) + \cdots + \mathbf{w}_N \nabla_N \sigma(\mathbf{w}),\end{aligned}\quad (18)$$

$\mathbf{w}_n \nabla_n \sigma(\mathbf{w})$  ( $n = 1, \dots, N$ ) is interpreted as the contribution of the  $n$ -th asset on the total risk of the portfolio and is called the **TRC (Total Risk Contribution)**.

# Risk Parity Portfolio i

A portfolio that satisfies

$$w_1 \nabla_1 \sigma(w) = \dots = w_N \nabla_N \sigma(w), \quad (19)$$

is called the **risk parity portfolio**. Dividing both sides of (18) with  $\sigma(w)$ , we have

$$1 = \frac{w_1 \nabla_1 \sigma(w)}{\sigma(w)} + \dots + \frac{w_N \nabla_N \sigma(w)}{\sigma(w)}.$$

Under (19),

$$\frac{w_n \nabla_n \sigma(w)}{\sigma(w)} = \frac{1}{N}, \quad (n = 1, \dots, N).$$

## Risk Parity Portfolio ii

$w^{RP}$  is the solution of the system of non-linear equations:

### Risk parity portfolio

$$\Sigma w^{RP} = \frac{\kappa}{w^{RP}} = \begin{bmatrix} \frac{\kappa}{w_1^{RP}} \\ \vdots \\ \frac{\kappa}{w_N^{RP}} \end{bmatrix}, \quad (20)$$
$$1^\top w^{RP} = 1.$$

(20) has no closed-form solution, but can be solved numerically.

# Properties

1. If all correlation coefficients are equal,

$$w_n^{RP} = \frac{\sigma_n^{-1}}{\sum_{n=1}^N \sigma_n^{-1}}, \quad (n = 1, \dots, N),$$

that is,

$$w^{RP} = \frac{\sigma^{-1}}{\mathbf{1}^\top \sigma^{-1}}, \quad \sigma^{-1} = \begin{bmatrix} \sigma_1^{-1} \\ \vdots \\ \sigma_N^{-1} \end{bmatrix}.$$

2. Suppose  $\sigma_1^2 = \dots = \sigma_n^2$  and the correlation coefficient between any pair of assets is constant. Then the risk parity portfolio is equivalent to the  $\mathbf{1}/N$  portfolio.

# Maximum Diversification Portfolio

The **maximum diversification portfolio** is the solution of

$$\max_w \frac{\sigma^\top w}{\sqrt{w^\top \Sigma w}}. \quad (21)$$

The optimal  $w$  in (21) is given by

## Maximum diversification portfolio

$$w^{MD} = \frac{1}{\iota^\top \Sigma^{-1} \sigma} \Sigma^{-1} \sigma. \quad (22)$$

(22) equalizes the risk allocation as

$$\frac{\nabla_1 \sigma(w^{MD})}{\sigma_1} = \dots = \frac{\nabla_N \sigma(w^{MD})}{\sigma_N}. \quad (23)$$

# Properties

1. If all correlation coefficients are equal,

$$w_n^{MD} = \frac{\sigma_n^{-1}}{\sum_{n=1}^N \sigma_n^{-1}}, \quad (n = 1, \dots, N).$$

2. Suppose  $\sigma_1^2 = \dots = \sigma_n^2$  and the correlation coefficient between any pair of assets is constant. Then the risk parity portfolio is equivalent to the  $\mathbf{1}/N$  portfolio.

# Passive Management Vs. Active Management

So far we have reviewed how to manage our portfolio in terms of the balance between the expected return and the risk (the variance or the expected shortfall). This style of portfolio management is called **active management**. Active management also involves discretionary selection of assets.

**Passive management** of a portfolio, on the other hand, is a investment strategy in which an investor tries to mimic a benchmark index. Passive management funds that mimic indices are called **index funds**. As the benchmark portfolio, index funds use stock indices, bond indices, currencies, commodities, or even hedge funds.

The goal in passive management is to minimize a discrepancy between a portfolio and the benchmark index.

# Tracking Error

Let  $y_t$  denote the return on the benchmark index at time  $t$ ,  $r_{nt}$  denote the return on asset  $n$  ( $n = 1, \dots, N$ ) at time  $t$  ( $t = 1, \dots, T$ ), and  $w_n$  denote the allocation weight for asset  $n$ . Then a discrepancy between a portfolio and the benchmark index at time  $t$  is given by

$$\begin{aligned} e_t &= y_t - \sum_{n=1}^N w_n r_{nt} = y_t - \begin{bmatrix} r_{1t} & \cdots & r_{Nt} \end{bmatrix} \begin{bmatrix} w_1 \\ \vdots \\ w_N \end{bmatrix} \\ &= y_t - r_t^T w. \end{aligned} \quad (24)$$

This discrepancy is called a **tracking error**.



# Tracking Error Minimization

A tracking error minimization problem is formulated as

## Tracking error minimization

$$\begin{aligned} \min_w \quad & \frac{1}{T} \sum_{t=1}^T (y_t - r_t^\top w)^2, \\ \text{subject to} \quad & \sum_{n=1}^N w_n = 1, \quad w_1 \geq 0, \dots, w_N \geq 0. \end{aligned} \tag{25}$$

# Tracking Error Minimization (Matrix Form)

By defining

$$\mathbf{y} = \begin{bmatrix} y_1 \\ \vdots \\ y_T \end{bmatrix}, \quad \mathbf{R} = \begin{bmatrix} r_1^\top \\ \vdots \\ r_T^\top \end{bmatrix}, \quad \mathbf{e} = \begin{bmatrix} e_1 \\ \vdots \\ e_T \end{bmatrix}.$$

Then the tracking error minimization problem is defined as

## Tracking error minimization (matrix form)

$$\begin{aligned} \min_w \quad & \frac{1}{T} \mathbf{e}^\top \mathbf{e}, \\ \text{subject to} \quad & \mathbf{e} = \mathbf{y} - \mathbf{R}\mathbf{w}, \quad \mathbf{w}^\top \boldsymbol{\iota} = 1, \quad \mathbf{w} \geq 0. \end{aligned} \tag{26}$$