

# portfolio\_construction\_with\_bond&stock

February 8, 2021

## 0.1 1. Import packages

```
[32]: import numpy as np
      from scipy.optimize import linprog
```

## 0.2 2. set parameters

```
[10]: # Get Prices and future prices, each has 1/3 probability
      stock_price = 20
      stock_prices_future = np.transpose(np.array([[40,20,12]]))
      bond_price = 90
      bond_prices_future = np.transpose(np.array([[100,100,100]]))
```

## 0.3 3. calculation

```
[45]: # Get expected profits
      expected_profit_stock = np.mean(stock_prices_future-stock_price)
      expected_profit_bond = 10

      # get returns for stock and bond
      stock_returns = (stock_prices_future - stock_price)/stock_price;
      bond_returns = (bond_prices_future - bond_price)/bond_price;

      # expected returns for stock
      expected_stock_return = np.mean(stock_returns)
      expected_bond_return = np.mean(bond_returns)
      print('expected_stock_return: ',expected_stock_return)
      print('expected_bond_return: ',expected_bond_return)
```

```
expected_stock_return:  0.19999999999999998
expected_bond_return:  0.11111111111111111
```

## 0.4 4.solve linear program maximizing expected profit

```
[46]: # define the function to maximize
      # why negative here? Because linprog is used to minimize
      f = [-1*expected_profit_stock, -1*expected_profit_bond]
```

```
A = [[20, 90]]
b = [50000]
x0_bounds = (0, None)
x1_bounds = (0, None)
result = linprog(f, A_ub=A, b_ub=b, bounds=[x0_bounds, x1_bounds])
```

```
[47]: print(result)
```

```
con: array([], dtype=float64)
fun: -9999.999985319324
message: 'Optimization terminated successfully.'
nit: 5
slack: array([6.92499598e-05])
status: 0
success: True
x: array([2.5000000e+03, 1.0383546e-07])
```

## 0.5 5. More Computation for cost, profit, risk and sharpe

```
[50]: # what is expected profit?
expected_profit = -np.dot(np.array(f), np.array(result.x));
expected_profit
```

```
[50]: 9999.999985319324
```

```
[59]: # compute cost of portfolio
cost = np.dot(np.array([stock_price, bond_price]), np.array(result.x))
print(cost)
```

```
49999.99993075004
```

```
[63]: np.squeeze(bond_price_future)
```

```
[63]: array([100, 100, 100])
```

```
[66]: # computation of Risk

portfolio_prices_future = np.dot(np.transpose(np.array([stock_price_future, np.
→squeeze(bond_price_future)])), np.array(result.x))
print('portfolio_prices_future: ', portfolio_prices_future)

portfolio_returns = (portfolio_prices_future - cost)/cost
print('portfolio_returns: ', portfolio_returns)
```

```
portfolio_prices_future: [99999.99985319 49999.99993179 29999.99996323]
portfolio_returns: [ 1.00000000e+00  2.07671838e-11 -4.00000000e-01]
```

**0.5.1** In python np.std use n as denominator not n-1, matlab std use n-1 that's where the difference!

```
[67]: # In python np.std use n as denominator not n-1, matlab std use n-1 that's ↵
      ↪where the difference!
      mu = np.mean(portfolio_returns)
      sigma = np.std(portfolio_returns)

      # calculate sharpe!
      sharpe = mu/sigma
      print('sharpe', sharpe)
```

sharpe 0.33968311027865006

**0.6** Here we should use n as denominator because the variance we calculate here is not an estimation

```
[89]: np.sqrt(np.sum((portfolio_returns-mu)**2/3))
```

[89]: 0.5887840576451439

```
[91]: sigma
```

[91]: 0.5887840576451439

**0.7** Why we should choose only stock? [2.5000000e+03, 1.0383546e-07], because stock has a higher return expected rate