2012 年度 大問 2

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1 問題

線形回帰モデル

2 解答

(1)

$$\hat{\beta} = (X^T X)^{-1} X^T y$$

(2)

$$E[\hat{\beta}] = \beta$$

(3)

$$\Sigma = \sigma^2 (X^T X)^{-1}$$

(4)

$$E[(y_{n+1} - x_{n+1}^T \beta)^2] = \sigma^2 (1 + x_{n+1}^T (X^T X)^{-1} x_{n+1})$$

3 おまけ

ソースコード 1 problem4

```
import numpy as np
2
   import matplotlib.pyplot as plt
   from tqdm import tqdm
   squared_errors = []
5
   answers = []
   for _ in tqdm(range(1000)):
7
8
       n = np.random.randint(1, 100)
       p = np.random.randint(1, 10)
       X = np.random.random((n, p))
10
       beta = np.random.random(p)
11
12
       sigma = np.random.random()
       epsilon = np.random.normal(0, sigma, n)
13
       y = X.dot(beta) + epsilon
       if np.linalg.matrix_rank(X) != p:
15
16
           continue
       beta_hat = np.linalg.inv(X.T.dot(X)).dot(X.T).dot(y)
       mse = 0
18
       for _ in range(100):
19
           x_np1 = np.random.random(p)
20
           y_np1 = x_np1.T.dot(beta) + np.random.normal(0, sigma)
21
22
           squared_error = (y_np1 - x_np1.T.dot(beta_hat)) ** 2
           mse += squared_error
23
       mse /= 100
24
       squared_errors.append(mse)
       answers.append(
26
           (sigma**2) * (1 + x_np1.T.dot(np.linalg.inv(X.T.dot(X)).
27
               dot(x_np1)))
       )
28
29
   plt.figure(figsize=(6, 6))
   plt.scatter(answers, squared_errors)
31
   plt.title("Simulated_Squared_Error_vs._Theoretical_Squared_Error"
32
   plt.xlabel("Theoretical_Squared_Error")
33
   plt.ylabel("Simulated_Squared_Error")
34
   plt.xscale("log")
  plt.yscale("log")
   plt.xlim(min(answers + squared_errors), max(answers +
      squared_errors))
   plt.ylim(min(answers + squared_errors), max(answers +
      squared_errors))
```

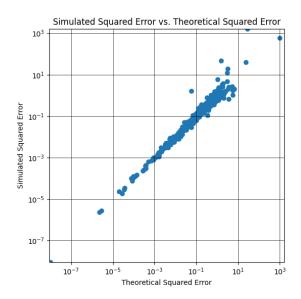


図 1 問題 4 のグラフ (シミュレーション値と理論値)