

1 (a)

$$t = w_0 + w_1 x_1 + \epsilon$$

$$\epsilon: N(0, \sigma^2 = 0.2^2) \Rightarrow \bar{\epsilon} = 0; \sigma_{\epsilon}^2 = 0.2^2; E(\epsilon^2) = 0.2^2$$

$$x_1: \text{Uniform}(-1, 1) \Rightarrow \bar{x}_1 = 0; \sigma_{x_1}^2 = \frac{1}{3}; E(x_1^2) = \frac{1}{3}$$

$$E(t) = E(w_0) + E(w_1 x_1) + E(\epsilon)$$

$$= w_0 + w_1 \cancel{E(x_1)} + \cancel{E(\epsilon)}$$

$$\boxed{E(t) = w_0}$$

$$\sigma_t^2 = E(t^2) - E(t)^2$$

$$t^2 = w_1^2 x_1^2 + 2w_0 w_1 x_1 + 2\epsilon w_1 x_1 + w_0^2 + 2\epsilon w_0 + \epsilon^2$$

$$E(t^2) = w_1^2 E(x_1^2) + 2w_1 \cancel{E(\epsilon x_1)} + w_0^2 + E(\epsilon^2)$$

$$= w_1^2 \frac{1}{3} + w_0^2 + 0.2^2$$

$$\boxed{\sigma_t^2 = w_1^2 \frac{1}{3} + 0.2^2}$$

$$\therefore \begin{cases} \bar{t} = w_0 \\ \sigma_t^2 = \left(\frac{w_1^2}{3} \right) + \sigma_{\epsilon}^2 \end{cases}$$

exact
soln

$$w_1^2 \sigma_x^2$$


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# SOLUTION HW 2
import numpy as np
import matplotlib.pyplot as plt
#-----
# DATA GENERATION
NROW = 2
NCOL = 2
NPTS = 40

# Exact coeff
w_exact = np.zeros((NROW))
w_exact = [-0.3, 0.5]

# Noise signal
e = np.zeros((NPTS))
e_std = 0.2
e_var = e_std**2
e_mean = 0.0
np.random.seed(1234)
e = np.random.normal(e_mean, e_std, NPTS)

# input basis vectors
x = np.zeros((NROW, NPTS))
x1 = np.random.uniform(-1.0, 1.0, NPTS)

for i in range(0, NPTS, 1):
    x[0][i] = 1.0
    x[1][i] = x1[i]

# output vector
t = np.zeros((NPTS))
for i in range(0, NPTS, 1):
    t[i] = e[i]
    for k in range(0, NCOL, 1):
        t[i] = t[i] + w_exact[k]*x[k][i]
#####
# part (a)
#####
t_mean_approx = np.sum(t)/NPTS
t_var_approx = np.sum(t*t)/NPTS - t_mean_approx**2
print('HW2 Part (a)')
print('approx mean(exact): ', t_mean_approx, '(' , w_exact[0], ')')
print('approx var (exact): ', t_var_approx, '(' , w_exact[1]**2/3 + e_var, ')')
#####
# part (b,c)
# MLE solution
#####
phi_trans = np.zeros((NROW, NPTS))
for i in range(0, NPTS, 1):
    phi_trans[0][i] = x[0][i]
    phi_trans[1][i] = x[1][i]

A = np.matmul(phi_trans, np.transpose(phi_trans))
B = np.matmul(phi_trans, t)
w_approx = np.linalg.solve(A, B)

print('HW2 Part (b)')
print('MLE approx (exact)', w_approx, '(' , w_exact, ')')
print('condition number A', "%4f" % np.linalg.cond(A))
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#####
x1_mean_approx = np.sum(x1)/NPTS
x1_var_approx = np.sum(x1*x1)/NPTS - x1_mean_approx**2
print('approx mean(var) x1: ', x1_mean_approx, '(' , x1_var_approx, ')')

e_mean_approx = np.sum(e)/NPTS
e_var_approx = np.sum(e*e)/NPTS - e_mean_approx**2
print('approx mean(var) e: ', e_mean_approx, '(' , e_var_approx, ')')

ez = np.sum((e-e_mean_approx)*(x1-x1_mean_approx))/NPTS
print('approx mean(var) ez: ', ez, '(' , 0, ')')

exit()
```


Department of Electrical and Computer Engineering
University of Massachusetts Lowell
EECE 5440 Computational Data Modeling I

Assignment #2

1. Consider the system

$$t = w_0 + w_1 x_1 + \varepsilon$$

where $w_0 = -0.3$ and $w_1 = 0.5$. The random variable ε is drawn from the normal distribution with mean 0 and variance 0.2^2 . The random variable x_1 is uniformly distributed between $(-1,1)$. Generate the observation t for 40 trials.

- a. Using the trial values for t find estimates for the mean and variance of t . Compare the approximate to the exact result.
- b. Using the observations for the j^{th} trial t_j and input $\underline{x}^{(j)} = [1, x_1^{(j)}]$ for $j = (0, 39)$. Find \underline{w} using the MLE approach .
- c. Compare the computed and exact result for $\underline{w} = [w_0, w_1]^T$.