

# Homework 1

## Simple Linear Regression

1

- **(a)** One of the target variable could be the amount of assets
- **(b)** It's continuous
- **(c)** The grades candidates provided in the material
- **(d)** I don't think it would be a linear model, instead it could be simulated by a curve

2

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In [16]: import numpy as np
from scipy.optimize import leastsq
import matplotlib.pyplot as plt
Xi=np.array([0,1,2,3,4])
Yi=np.array([0,2,3,8,17])
```

```
In [39]: xbar=np.mean(Xi)
ybar=np.mean(Yi)
print ('The answer of question (a) is')
print('xbar={0:1.2f},y={1:1.2f}'.format(xbar,ybar))
```

The answer of question (a) is  
xbar=2.00,y=6.00

```
In [40]: syy = np.mean((Yi-ybar)**2)
syx = np.mean((Yi-ybar)*(Xi-xbar))
sxx = np.mean((Xi-xbar)**2)
print ('The answer of question (b) is')
print ('Syy={0:1.2f},Syx={1:1.2f},Sxx={2:1.2f}'.format(syy,syx,sxx))
```

The answer of question (b) is  
Syy=37.20,Syx=8.00,Sxx=2.00

```
In [41]: betal = syx/sxx
beta0 = ybar - betal*xbar
print('The answer of question (c) is')
print('betal={0:1.2f},beta0={1:1.2f}'.format(betal,beta0))
```

The answer of question (c) is  
betal=4.00,beta0=-2.00

```
In [35]: x=2.5
y=betal*x+beta0
print('The answer of question (d) is')
print('y=',y)
```

The answer of question (d) is  
y= 8.0

$$3. (a) z(t) = z_0 e^{-at}$$

$$\text{Let } y(t) = \ln(z(t)).$$

$\therefore y(t) = \ln z_0 - at$ . which is a linear model.

$$4. (b). \quad y(t) = \ln z(t)$$

$$\bar{t} = \frac{1}{n} \sum t_i \quad \bar{y} = \frac{1}{n} \sum y_i$$

$$S_t^2 = \frac{1}{n} \sum (t_i - \bar{t})^2, \quad S_y^2 = \frac{1}{n} \sum (y_i - \bar{y})^2.$$

$$S_{ty} = \frac{1}{n} \sum (y_i - \bar{y})(t_i - \bar{t}).$$

$$\beta_1 = S_{ty} / S_t. \quad \beta_0 = \bar{y} - \beta_1 \bar{t}.$$

$$\Rightarrow \beta_1 = -a. \quad \beta_0 = \ln z_0.$$

$$z_0 = e^{\beta_0} \quad a = -\beta_1.$$



In [ ]:

```
x=t
y=np.log(z)
xm = np.mean(x)
ym = np.mean(y)
syy = np.mean((y-ym)**2)
syx = np.mean((y-ym)*(x-xm))
sxx = np.mean((x-xm)**2)
beta1 = syx/sxx
beta0 = ym - beta1*xm
alpha=-b1
z0=exp(b0)
```

$$4. (a) R_{SS}(\beta) = \sum_{i=1}^N (y_i - \beta x_i)^2.$$

$$(b) \frac{dR_{SS}(\beta)}{d\beta} = \sum_{i=1}^N 2(y_i - \beta x_i)(-x_i) = 0.$$

$$\sum_{i=1}^N -2y_i x_i + 2\beta x_i^2 = 0.$$

$$\beta \sum_{i=1}^N x_i^2 = \sum_{i=1}^N y_i x_i$$

$$\beta = \frac{\sum_i x_i y_i}{\sum_i x_i^2}.$$

