$$W = \frac{\sum_{i=1}^{n=3} \chi_i y_i}{\sum_{i=1}^{n=3} \chi_i \chi_i} = \frac{x_4 y_4 + x_5 y_5 + x_4 y_3}{x_4 x_4 + x_5 x_5 + x_5 x_3} = \frac{o+4+0}{o+1+4} = \frac{4}{5}$$

$$g(x) = W^{T} \phi(x) = [2, -1] \begin{bmatrix} x \\ x^{2} \end{bmatrix} = 2 x - x^{2}$$

3.

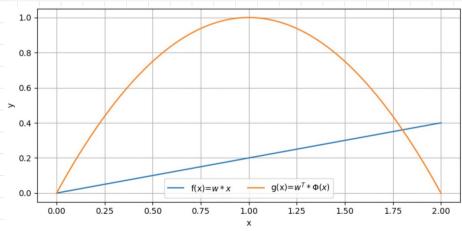
```
import numpy as np
import matplotlib.pyplot as plt

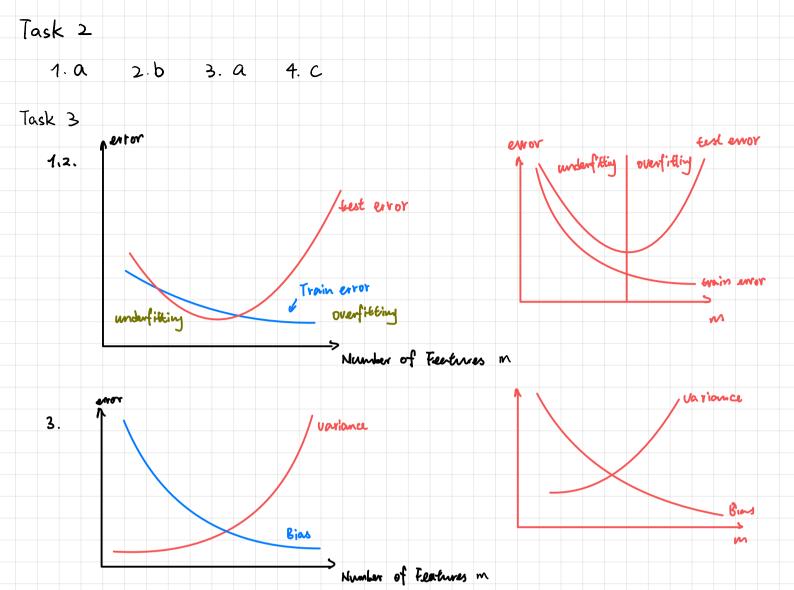
x = np.linspace(0, 2, 400)

def g_x(x):
    return 2*x-x**2

def f_x(x):
    return 0.2*x

plt.figure(figsize=(8, 4))
plt.xlabel("x")
plt.ylabel("y")
plt.ylabel('y')
#plt.scatter(datapoints, f_x(datapoints), label='Data Points')
plt.plot(x, f_x(x),'-', label='f(x)=$w*x$')
plt.plot(x, g_x(x),"-", label='g(x)=$w^T*\Phi(x)$')
plt.legend(fontsize=10, ncol=2)
plt.grid(True)
plt.tight_layout ()
plt.show()
```





U. a) the erain error will increase.

Becomie  $\mathcal{E}_{RR} = \sum_{i=1}^{N} (y_i - \hat{f}(x_i))^2 + \sum_{i=1}^{M} w_i^2$  is greater than  $\mathcal{E} = \sum_{i=1}^{N} (y_i - \hat{f}(x_i))^2$ ,

which means 3 \sum no onsermins the flexibility of the model

b) the test error will decrease

Because regularization will help to prevent overfitting duch 5 = Ni

e) the bias of I will increme

d) the variance of f will decreve

$$4. \times '=AX$$

= 
$$(A^T)^{-1}(XX^T)^{-1}A^{-1}AXY^T$$
 | Aist invertible

$$= (A^T)^{-1} (XX^T)^{-1} Y Y^T = (A^T)^{-1} \hat{W}_{ols}$$

$$\hat{Y}_{OLS} = \hat{w}_{OLS}^T X = ((XX^T)^{-1}Xy^T)^T X = YX^T(XX^T)^{-1}X$$

$$= y \chi^{\mathsf{T}} \cdot (\chi \chi^{\mathsf{T}})^{-1} \cdot ((A^{-1})^{\mathsf{T}})^{\mathsf{T}} \cdot A \chi$$

$$= y \times^T \cdot (x \times^T)^{-1} \cdot x = \hat{y}_{ols}$$

## a. A ist orthogonal Matrix => A7 = A-1

$$\hat{\mathbf{y}}_{RR} = \hat{\mathbf{N}}_{RR}^{\mathsf{T}} \mathbf{X} = \left[ (\mathbf{X} \mathbf{X}^{\mathsf{T}} + \mathbf{\lambda} \mathbf{I})^{-4} \mathbf{X} \mathbf{y}^{\mathsf{T}} \right]^{\mathsf{T}} \mathbf{X} = \mathbf{y} \mathbf{X}^{\mathsf{T}} \left( (\mathbf{X} \mathbf{X}^{\mathsf{T}} + \mathbf{\lambda} \mathbf{I})^{-4} \right)^{\mathsf{T}} \mathbf{X}$$

$$\hat{y}_{RR} = y(AX)^T ((AXX^TA^T + \lambda I)^{-1})^T AX$$

$$= \sqrt{\chi^{7}[(\chi\chi^{7} + \lambda I)^{-3}]} \chi = \hat{\chi}_{RR}$$

1. reject

This paper does model evaluation and model selection at the same time.

They could be too optimistic

2. for exosumple, we could regard  $K_A, K_2, K_3, K_8, K_5$  as training set, and  $K_6$  as test set

i.e. label: 
$$4 - 4 + 4 - \frac{x_1}{x_2}$$
,  $x_2$ ,  $x_3$ ,  $x_4$ ,  $x_5$ ,  $x_6$ 

train set

pred:  $e \in A$ 

the numbers of positive and negative ensamples are same size. In this case the train set has more exemples with + label, therefore it predict 166 would be + , meanwhile the test set has more ensamples with - label because positive and negative ensamples both have 100 ensamples.