

1a) Let $w \circ b$ be the element-wise product of the vector w and b . the results of that vector would be $[w_1b_1, w_2b_2, w_3b_3, \dots, w_nb_n]$. $a^T (w \circ b) = a_1w_1b_1 + a_2w_2b_2 + a_3w_3b_3 + \dots + a_nw_nb_n$

To calculate the gradient with respect to $w([d/w_1, d/w_2, d/w_3, \dots, d/w_n])$, the gradient would be $[a_1b_1, a_2b_2, a_3b_3, \dots, a_nb_n]$ which is just $a \circ b$.

b) $f(w) = \text{tr}(Bw w^T A)$ which we will use ' instead t to represent transpose

$$\begin{aligned} \nabla_w f(w) &= \nabla_w \text{tr}(B * w * w' * A) \\ &= \text{tr}(\nabla_w (B * w * w' * A)) \\ &= \text{tr}((B * w * w' * A)' * \nabla_w (w * w')) \\ &= \text{tr}((A' * B' * w * w')' * \nabla_w (w * w')) \\ &= \text{tr}(w * w' * A' * B' * \nabla_w w) \\ &= \text{tr}(w' * (w' * A' * B') \nabla_w w) \\ &= w' * (A * B) * w + w' * (A' * B') * w' \\ &= (A * B * w) + (B' * A' * w) \end{aligned}$$

The answer will be $\nabla_w f(w) = A * B * w + B' * A' * w$

c) $f(w) = \text{tr}(Bw w^T A)$

To find the Hessian matrix, we need to take the second partial derivative based on the gradient $A * B * w + B' * A' * w$. In this case it would be

$$\begin{aligned} H &= \nabla_w (A * B * w) + (B' * A' * w) \\ H &= A * B + B' * A' \end{aligned}$$

This will be the Hessian matrix for $f(w)$

D) In order to see if $f(w)$ is a convex function, the eigenvalue of the Hessian matrix needs to be positive.

```

1 import numpy as np
2
3 A = np.array([[2, -2], [-2, 3]])
4 B = np.array([[2, -1], [-1, 2]])
5 H = np.matmul(A,B)+np.matmul(np.transpose(B),np.transpose(A))
6 print(H)
7 eigenvalues= np.linalg.eigvals(H)
8 print(eigenvalues)

```

PROBLEMS OUTPUT DEBUG CONSOLE TERMINAL

```

PS C:\Users\andre\OneDrive\Desktop\Uni\Year4\AI 2\Assignment1> python .\part_d.py
[[ 12 -13]
 [-13  16]]
[ 0.84705356 27.15294644]
PS C:\Users\andre\OneDrive\Desktop\Uni\Year4\AI 2\Assignment1> 

```

The eigen value return [0.84,27.15] and since they are both positive, it means the $f(w)$ is a convex function.

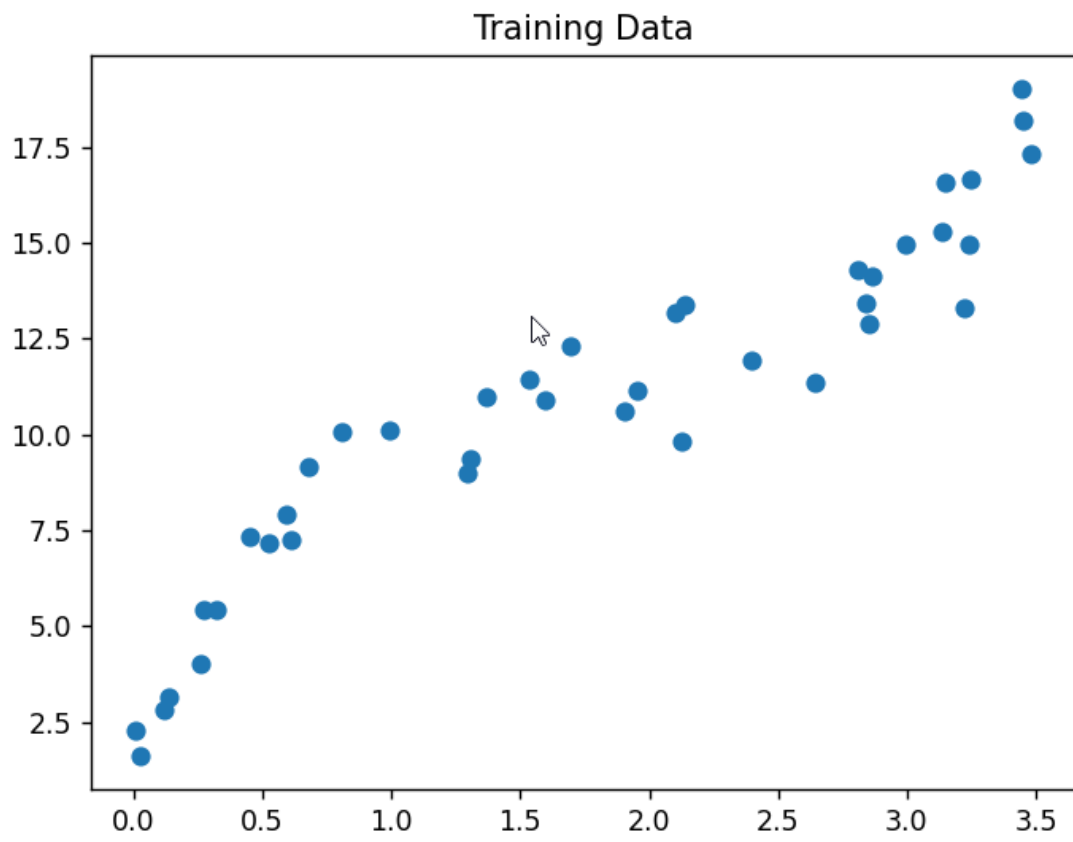
$$\begin{aligned}
 e) f(w) &= \log(\sigma(w^T x)) = -\log(1 + e^{-(w^T x)}) \\
 &= -\frac{d}{dw} \log(1 + e^{-(w^T x)}) \\
 &= -\frac{1}{(1 + e^{-(w^T x)})} \frac{d}{dw} (e^{-(w^T x)}) \\
 &= -\frac{1}{(1 + e^{-(w^T x)})} * (e^{-(w^T x)}) \frac{d}{dw} (-w^T x) \\
 &= \frac{1}{(1 + e^{-(w^T x)})} * (e^{-(w^T x)}) * x \\
 &= \frac{x e^{-(w^T x)}}{(1 + e^{-(w^T x)})}
 \end{aligned}$$

Which would be the gradient for $f(w)$

Question 2:

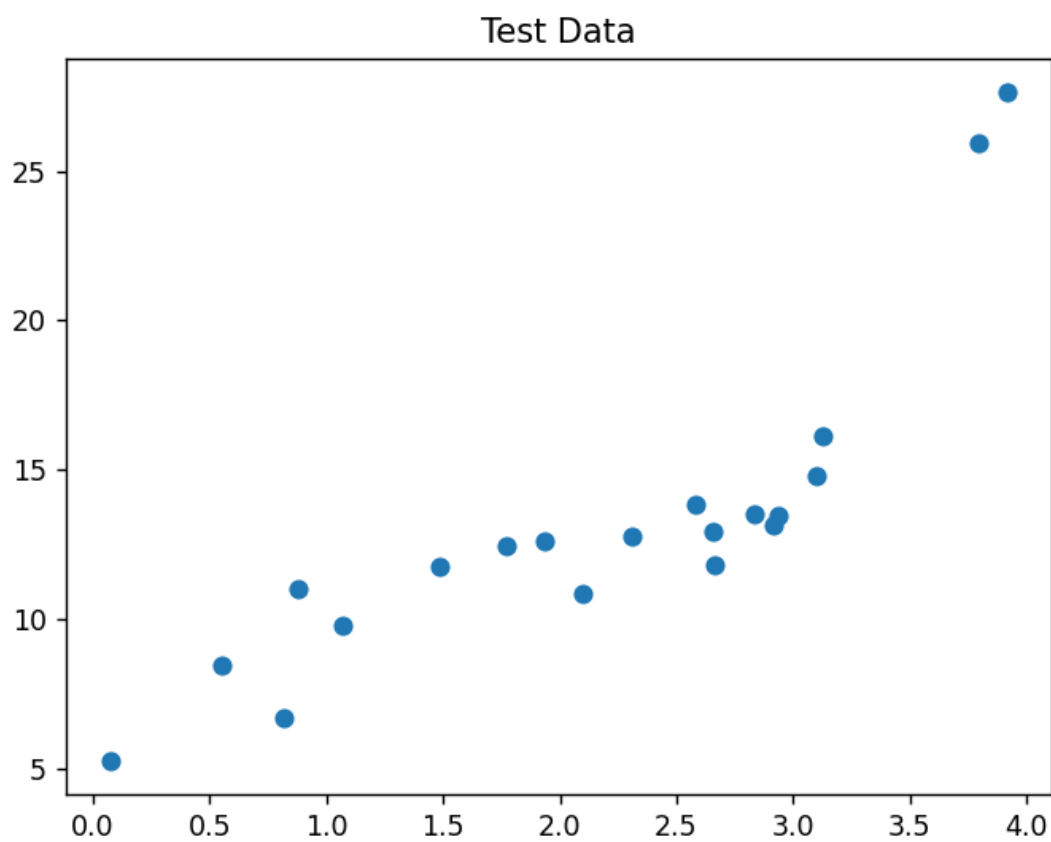
a)

Figure 1

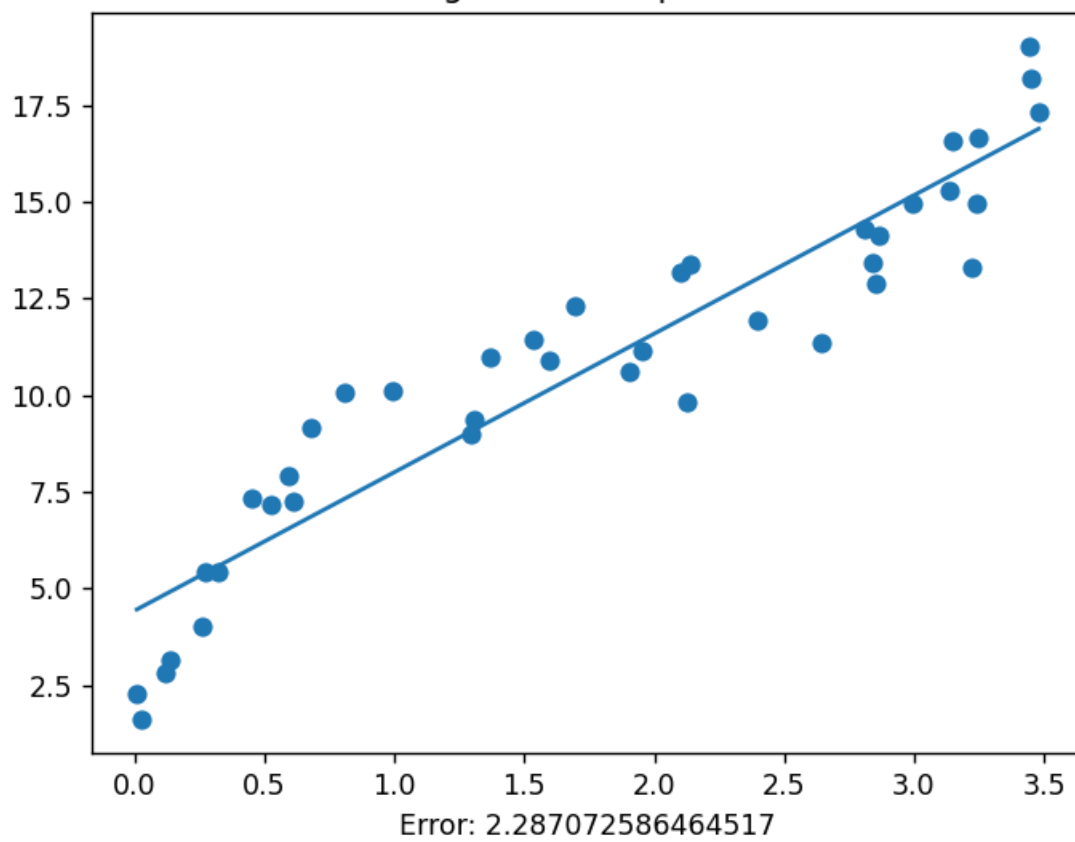


x=1.319 y=13.63

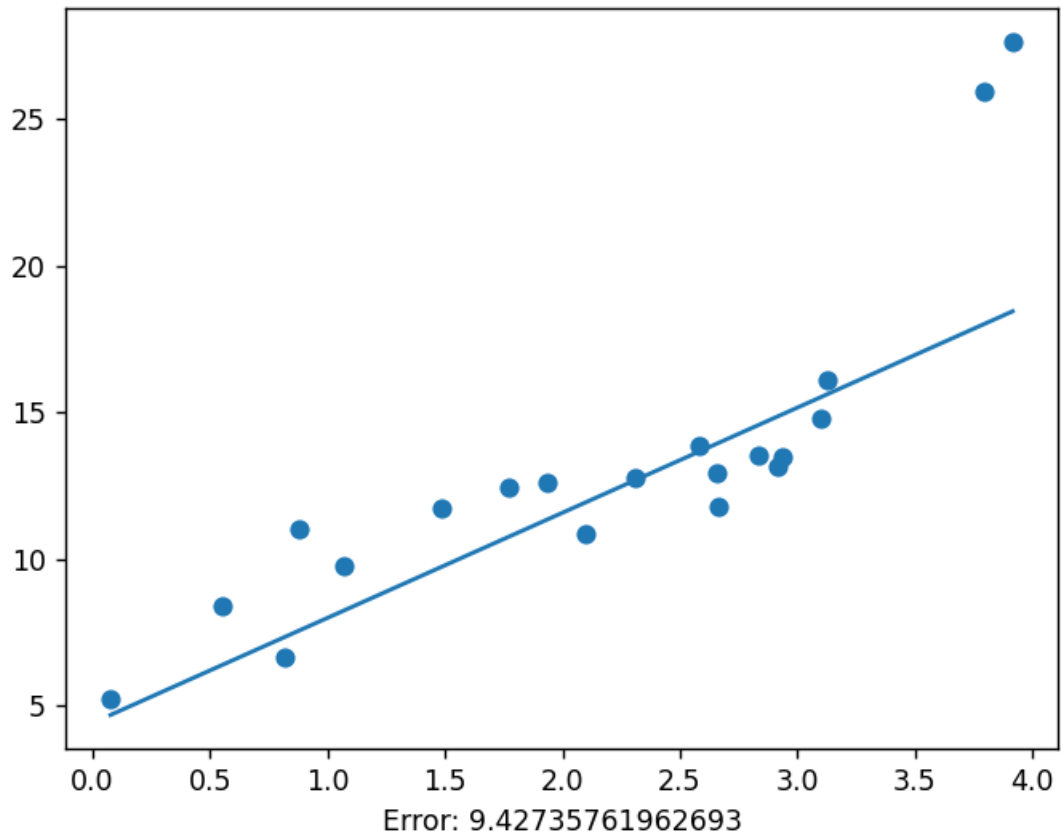
Figure 1



Training model for question 2B

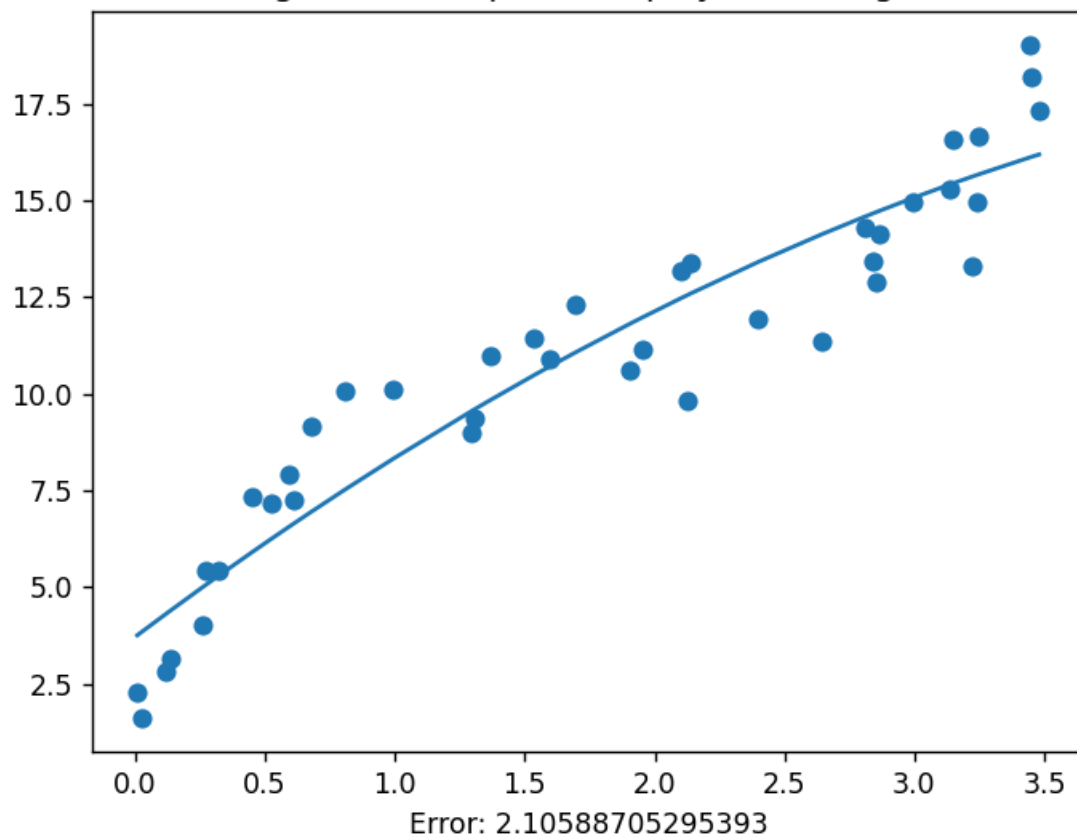


Testing model for question 2C

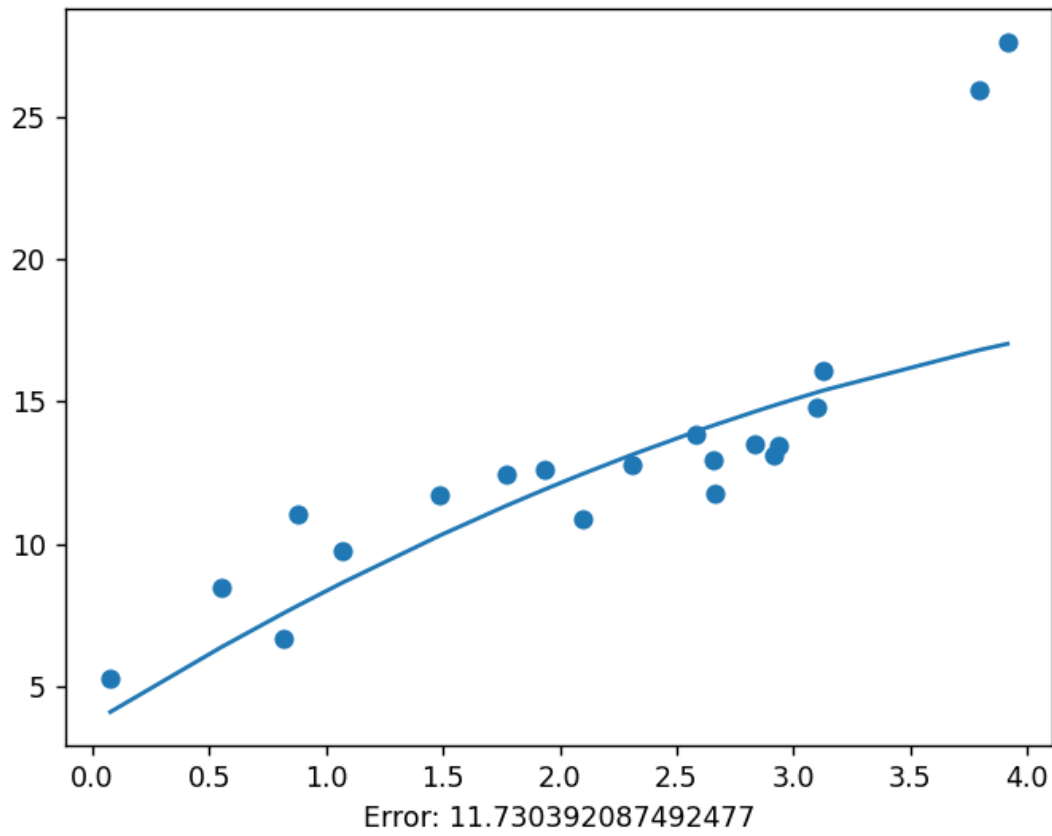




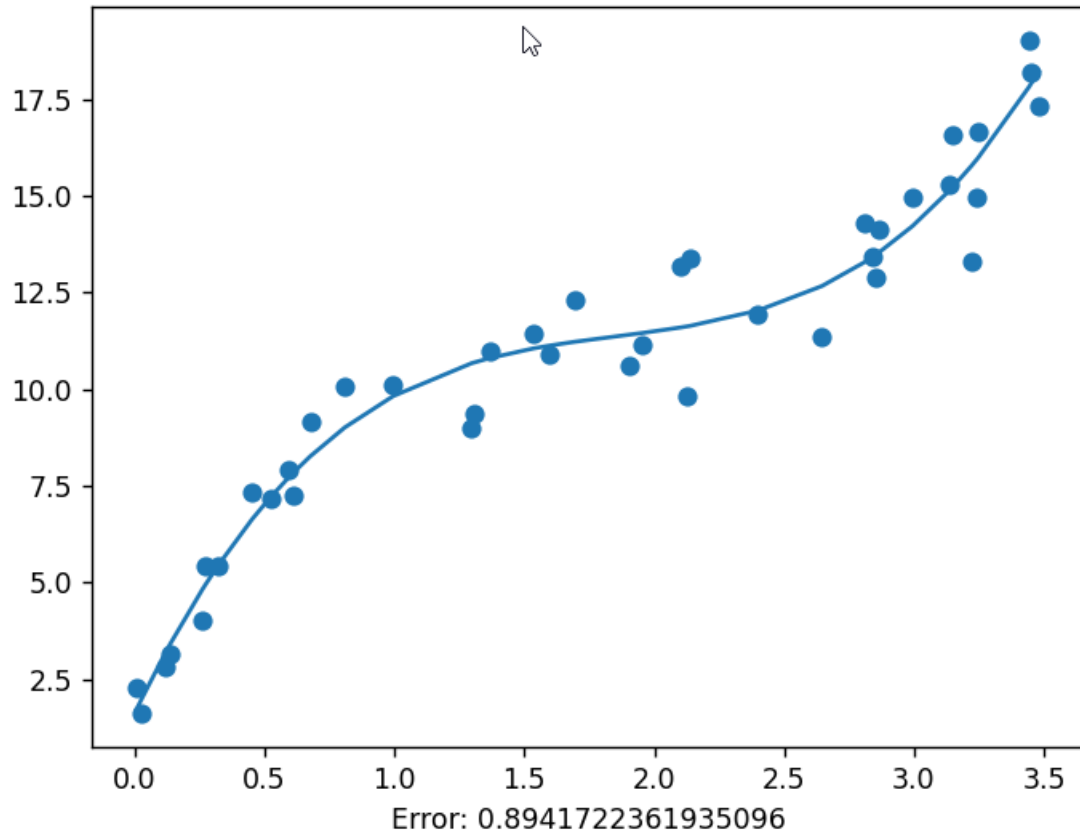
Training model for quadratic polynomial regression



Test model for quadartic polynomial regression

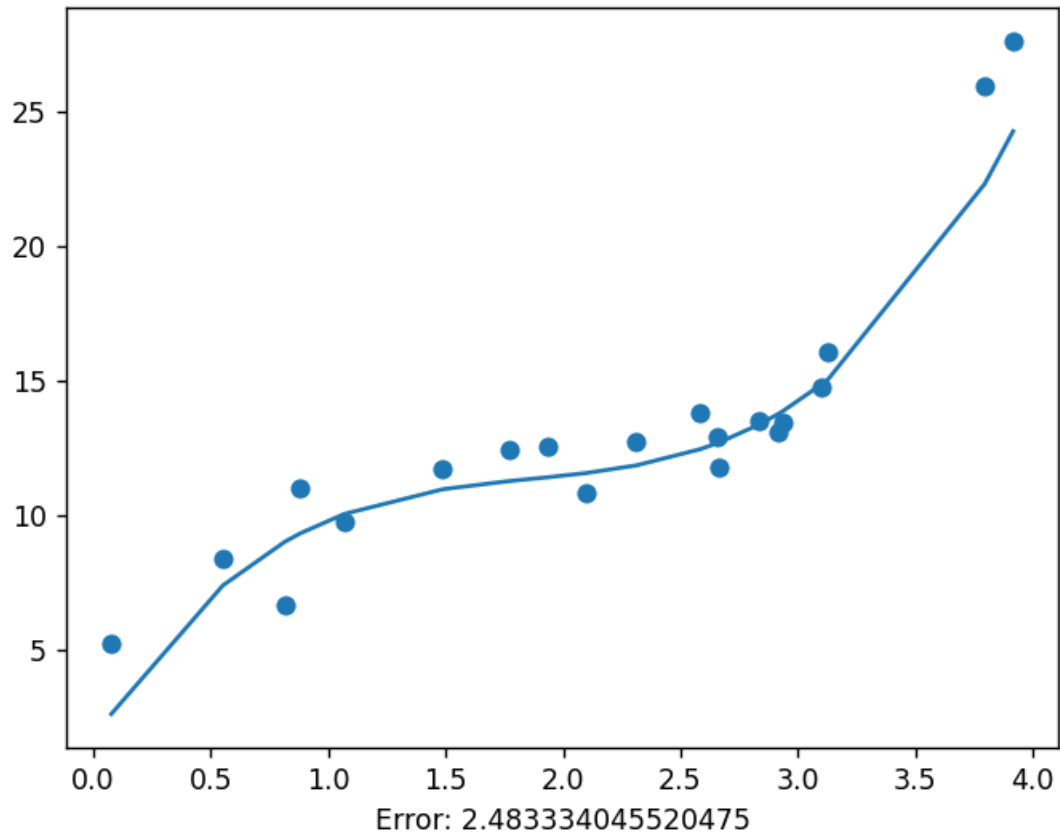


Training model for cubic polynomial regression

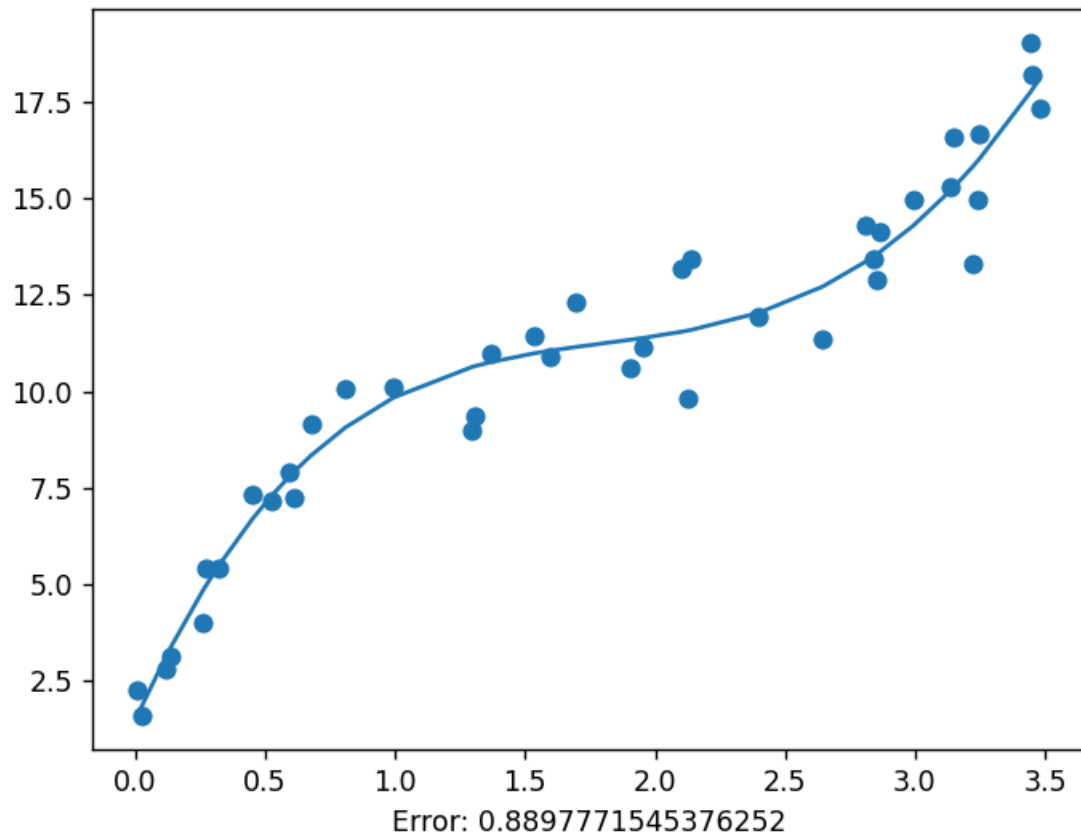


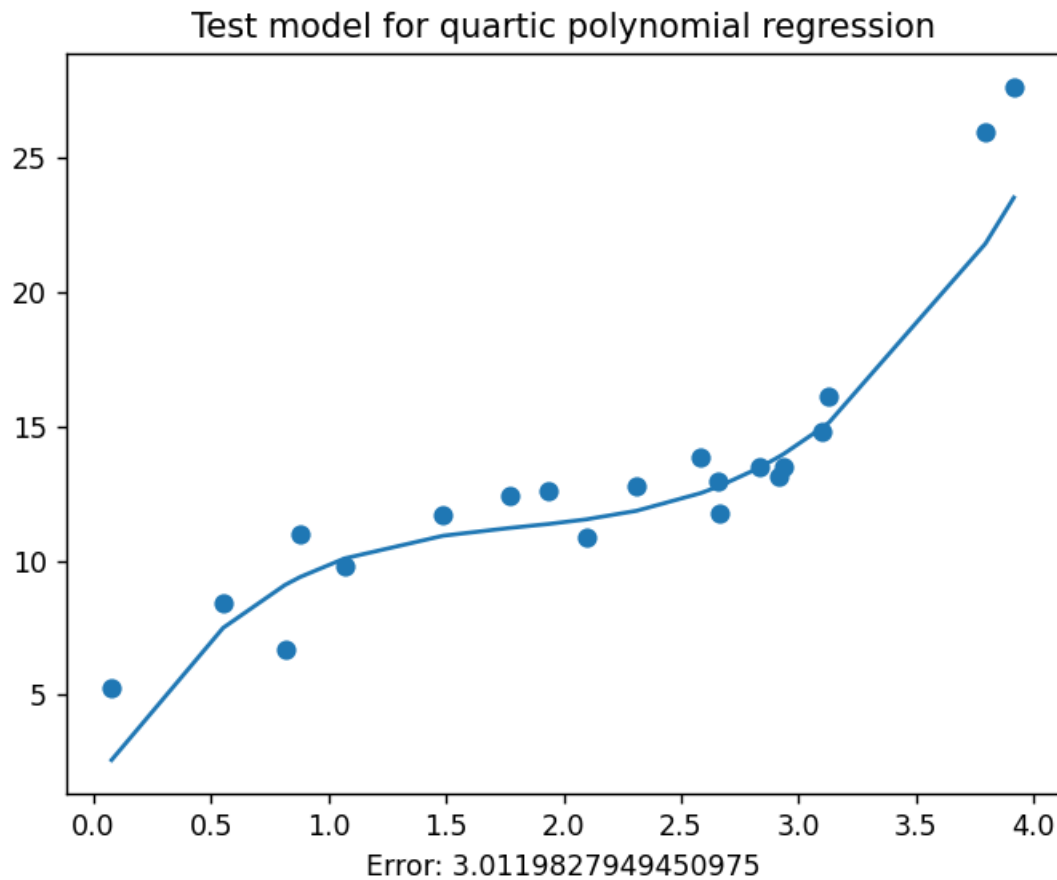
WS

Test model for cubic polynomial regression



Training model for quartic polynomial regression





Out of all of different degree of polynomial regression listed above, the best fit will be the one with the smallest error value for the testing and training mode. The best equation in the equations above is cubic followed by quartic, linear and quadratic. Even though the quartic has a lower training error than cubic, the actual test error is 0.6 higher than the cubic one which means overfitting.

Question 3A

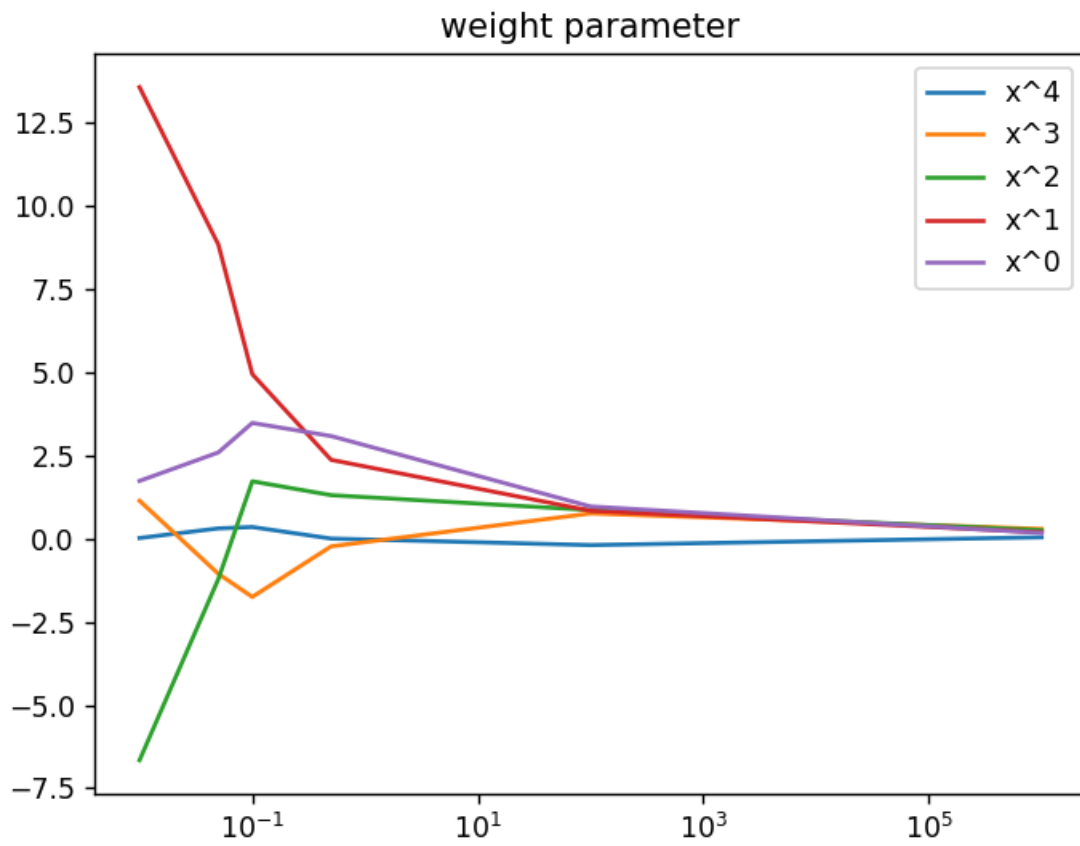
λ	Training Error	Test Error
0.01	0.897482895514694	2.5058009642404486
0.05	1.032312555684975	1.4947374059092655
0.1	1.366437903365513,	2.223431462310436
0.5	3.8514214385291874	13.230161041169646
100	13.948781079351482	25.293945460628255
10^6	32.82941976048144	34.44555027735635

The best λ for fitting the data would be 0.05 even though 0.01 had a lower training error than 0.05, the test error is a lot higher as well. Summing both errors will result in 0.05 having an overall lower error rate.



Question 3B

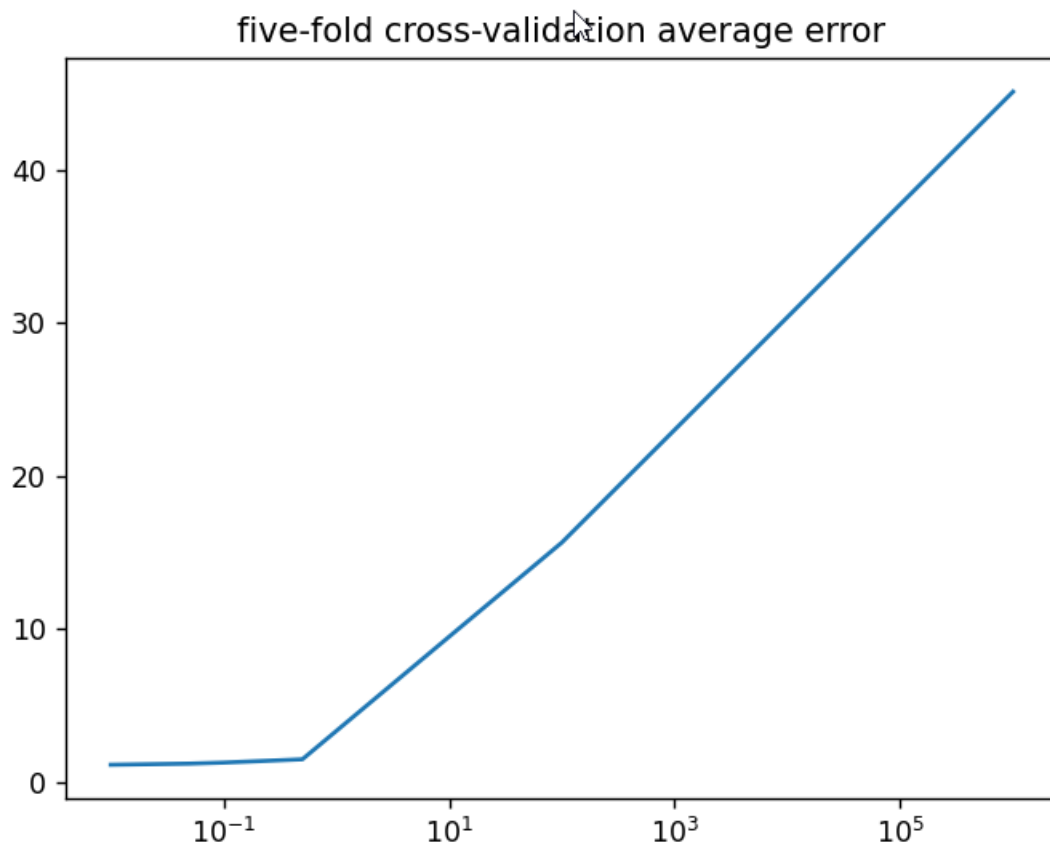
	$\lambda = 0.01$	$\lambda = 0.05$	$\lambda = 0.1$	$\lambda = 0.5$	$\lambda = 100$	$\lambda = 10^6$
W0	1.432e-02	3.035e-01	3.4491e-01	-2.852e-03	-1.975e-01	3.631e-02
W1	1.135	-1.059	-1.754	-2.360e-01	7.524e-01	2.865e-01
W2	6.659	-1.216	1.718	1.301	8.557e-01	2.490e-01
W3	1.355e+01	8.828	4.936	2.363	8.334e-01	1.864e-01
W4	1.729	2.584	3.470	3.076	9.618e-01	1.667e-01



Question 3C

	$\lambda = 0.01$	$\lambda = 0.05$	$\lambda = 0.1$	$\lambda = 0.5$	$\lambda = 100$	$\lambda = 10^6$
Average test error	1.154	1.227	1.294	1.509	15.667	45.101

Figure 1



The best average error is $\lambda = 0.01$ which is different from 3a but 0.01 had the lowest training error in 3a so the five-fold cross-validation can determine the best regularization parameter

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