(2-1.2) ()ii). [Plots were on the previous page].

Sigma basically represents the neighborhow,
that we take around our target
point. So, a larger sigma implies a
bigger neighbourhood.

For a sigma=1, the neighborhood is
really small, and as we can see the
model is arefitting on the data (High
For sigma=100 the neighborhood is
really large and the model is underfitting on the data (High bias).

For sigma=10, We observe that the
model fits just perfectly to the data.

-> Variation of hambde.

Higher the Sambda, more will be the regularisation penalty, and the variance of the model will decrease (more bias).

At dambda = 0.1, the model fits the data just perfectly, but as we increase lambda, the regularisation penalty forces the model to clevrese its degree of freedom, and hence, for righer lamba's the curve that is fit, by our model is more that is fit, by our model is more and linear lambdas.

Q-2.1]. Given K (x, x') is valid kernel. Then there has to exist &: RM >H S.T K(x, x')= (x) (x) - () (i) Consider a function prew of Rm > H $\phi_{\text{new}}(x) = \phi(g(x)).$ $\phi_{\text{new}}(x)^{7} \phi_{\text{new}}(y) = \phi(g(x))^{7} \phi(g(y)).$ = K(g(n), g(y))Hence, K(y(x), g(y)) can be represented in terms of imner space product of some frew. Hence knew (x,y)= K(g(x), g(y)) is a valid (ii) det $q(x) = \frac{1}{40} a_i x^2$ Property-1: If the K, K, are valid kernely.

Then X, K, + X, K, are valid kernely.

H & A & B O (Roovedin class]

from then Kink, is also a valid kerney (proved in class). property 3: - Al K, is a valid kernel then (k,) is a valid kernel $\forall d \in N$. (Proved in part 1). Using property 1,3 we can say that for any i E {0,1, -- n-1}. Hence sumpdation over all i EN, i < n Z a; h(x,x') is also a valid kernel Mence proved., q (k(n, n')) is also a valid
kernet (Q. 2.2]

My Kernel = K(x,y) = (1 + x + y) 4

This gives an error < 7000 which is

desired. [Plot attached on the next page]