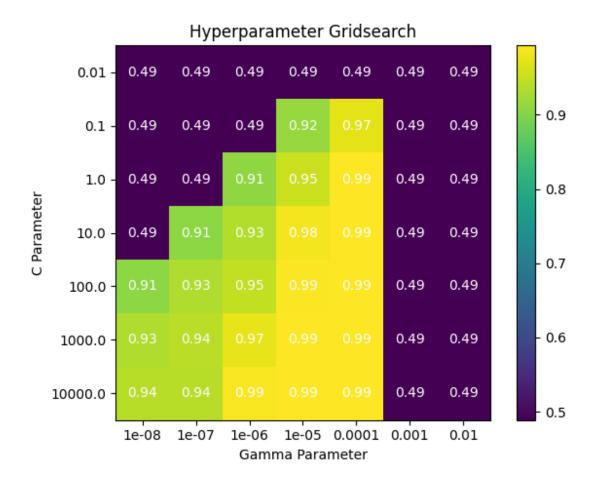
NCTU Introduction to Machine Learning, Homework 4 Part. 1, Coding (50%):

Q1

No output

Q2

Q3



Part. 2, Questions (50%):

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1. For a matrix to be positive semidefinite, all of its eigenvalues are non-negative. We already know that K is symmetric, K = V \wedge V, where V is an orthonormal workin Vt and the diagonal motrix \Lambda contains the eigenvalues \Lambda_t of K. In order to be a valid kernel, the elgenvalues must all be positive so that \Phi: X_1 = (\sqrt{N_t} \vee V_t)^2_{t_1} \in \mathbb{R}^n \Rightarrow \Phi(X_1) \Phi(X_2) = \frac{n}{2} \times V_t \vee V_t
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Kernel Furction. 3. Suppose we already know that given valid kernels $K_1(x,x')$ and $K_2(x,x')$, the following K(x,x') will also be valid; $K(X,X') = k_1(x_1x') + k_2(x_1X') - - - 0$ k(x,x1) = k1(x,x1)k2(x,x1) --- @ k(x,x') = exp(k,(x,x)) - - - 3 K(x,x') = f(x) K, (x,x') f(x') - - - @ (a) k(x,x1) = k,(x,x1)+1 Sympose that there exist K2(XXX)=1, K2 will be a NXN all ones matrix, where N is the number of data. Since the eigenvalue of K2 MUST be N and O, Kz is positive semidefinite, which means kz(x,x1)=1 is valid, K(x,x') = K1(x,x') + 1 - K1(x,x') + K2(x,x') By @, k(x,x') is a valid kernel function to (b) Suppose that ki(xix) = (xTx) 2 and M=[0], N=[0] k,(m,n)=(m7n)2=0 K1 (m, m) = 1 k, (n, n) = 1 3 K1=[0] Since we know that k(x,x) = k(m,n)-1 K=[0-1] and $\lambda=1,-1 \Rightarrow$ not positive semidefinite, not valid

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(c) k(x,x') = k_1(x_1x')^2 + e^{||x||^2}
= k_1(x_1x')^2 + f(x) k_2(x_1x') f(x'), where f(x) = e^{||x||^2}
f(x') = e^{||x'||^2}

By ② we know that k_1(x_1x')^2 is valid

By ③ we see f(x) k_2(x_1x') = 1 is valid

And by ③ we conclude that k(x_1x') is valid

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e^{x} = 1 + \frac{x_1}{1!} + \frac{x_2}{2!} + \cdots
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Therefore e^{x_1(x_1x')} = \frac{k_1(x_1x')}{1!} + \frac{k_1(x_1x')^2}{2!} + \frac{k_1(x_1x')^3}{3!} + \cdots

By ⑤ ③ and mathematical induction we know that k_3(x_1x') = e^{k_1(x_1x')} is a valid kernel.

By ⑥ , k(x_1x') = k_1(x_1x')^2 + e^{k_1(x_1x')} = k_1(x_1x')^2 + k_3(x_1x') is valid

Explicitly k_1(x_1x') = k_1(x_1x')^2 + k_1(x_1x')^2 + k_3(x_1x') is valid

Explicitly k_1(x_1x') = k_1(x_1x')^2 + k_1(x_1x')^2 + k_3(x_1x') is valid

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Explicitly k_1(x_1x') = k_1(x_1x')^2 + k_1(x_1x')^2 + k_1(x_1x')^2 + k_2(x_1x') is valid

Explicitly k_1(x_1x') = k_1(x_1x')^2 + k_1(x_1x')^2 + k_2(x_1x') is valid

Explicitly k_1(x_1x') = k_1(x_1x')^2 + k_2(x_1x') = k_1(x_1x')^2 + k_2(x_1x') is valid

Explicitly k_1(x_1x') = k_1(x_1x')^2 + k_2(x_1x') = k_1(x_1x')^2 + k_2(x_1x') = k
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4.
$$L(x, 0) = (x-2)^2 - \alpha(3 - (x+3)(x-1))$$
, $\alpha \ge 0$
 $L(\alpha, \alpha) = x^2 + 4x + 4 - \alpha(-x^2 - 2x + 6)$, $\alpha \ge 0$
 $\frac{\partial L(x, \alpha)}{\partial x} = 0 \Rightarrow 0 = 2x - 4 + 2\alpha x + 2\alpha \Rightarrow x(2+2\alpha) = 4-2\alpha \Rightarrow x = \frac{2-\alpha}{1+\alpha}$
Dual problem
 $\Rightarrow \text{maximize } (\frac{2-\alpha}{1+\alpha} - 2)^2 - \alpha(-(\frac{2-\alpha}{1+\alpha})^2 - 2(\frac{2-\alpha}{1+\alpha}) + 6)$
subject to $\alpha \ge 0$