

## CS446: Machine Learning, Fall 2018, Homework 2

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### Problem (2)

(a)

Since  $A$  is symmetric,  $A$  is unitarily diagonalizable, meaning that

$$A = P^T D P$$

Where  $P^T = P^{-1}$  and  $D$  is diagonal

Since  $D$  is diagonal and  $A$  positive semi-definite, the eigenvalues of  $A$  occupy the diagonal entries of  $D$  and all eigenvalues are non-negative. As a consequence, we can define a square root for  $D$ , which is the matrix obtained by taking the square root of each diagonal entry in  $D$ . We call this matrix  $E$

$$\begin{aligned} A &= P^T E E P \\ \implies x^T A x' &= x^T P^T E E P x' \\ &= (E P x)^T (E P x') \end{aligned}$$

Thus, we see that a feature transformation  $\phi$  exists defined by  $\phi(x) = E P x$  such that  $x^T A x' = k(x, x') \phi(x)^T \phi(x')$ .

(b)

Since  $k$  is a valid kernel,  $k(x, x')$  can be decomposed into the inner product of some feature transformation  $\phi$ . That is,  $k(x, x') = \phi(x)^T \phi(x')$ .

Define a new feature transformation  $\psi(x) = f(x) \phi(x)$ . Then observe that

$$\psi(x)^T \psi(x^*) = f(x) \phi(x)^T \phi(x^*) f(x^*)$$

(c)

We show that  $x^T K x \geq 0$  for all  $x \in \mathbb{R}^n$ . Recall that inner products produce non-negative values in  $\mathbb{R}$  and that they are symmetric. Thus  $K$  is a symmetric matrix with no negative entries. It follows that

$$\begin{aligned}
 x^T A x &= \sum_{i,j} x_i x_j A_{ij} \\
 &= 2 \sum_{i,j>i} x_i x_j A_{ij}
 \end{aligned}$$

Since  $x_i x_j A_{ij} = x_j x_i A_{ij}$

Now suppose that arbitrary  $x$  is given.  $x$  can be decomposed as the sum of two vectors  $x_1$  and  $x_2$  such that every entry in  $x_1$  is non-negative and every entry in  $x_2$  is non-positive. It follows that

$$\begin{aligned}
 x^T A x &= (x_1 + x_2)^T A (x_1 + x_2) \\
 &= x_1^T A x_1 + x_1^T A x_2 + x_2^T A x_1 + x_2^T A x_2
 \end{aligned}$$

From (1), we know that:

$$x_1^T A x_1 = 2 \sum_{i,j>i} x_1^i x_1^j A_{ij}$$

From how we defined  $x_1$ , we conclude that this foregoing expression is non-negative. By similar reasoning, we can conclude that  $x_2^T A x_2$  is non-negative

By construction,  $x_1^T A x_2$  and  $x_2^T A x_1$  are both zero, since wherever  $x_1$  is not zero,  $x_2$  is zero and vice versa. Hence

$$x_1^T A x_1 + x_1^T A x_2 + x_2^T A x_1 + x_2^T A x_2 \geq 0$$

This completes the proof and the problem.

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