

## Homework 2 Written Assignment

### **10-605/10-805: Machine Learning with Large Datasets**

**Due Tuesday, September 29th at 1:30:00 PM Eastern Time**

Submit your solutions via Gradescope, **with your solution to each subproblem on a separate page**, i.e., following the template below. Note that Homework 2 consists of two parts: this written assignment, and a programming assignment. The written part is worth **30%** of your total HW2 grade (programming part makes up the remaining 70%).

# 1 Nyström Method (30 points)

Nyström method. Define the following block representation of a kernel matrix:

$$\mathbf{K} = \begin{bmatrix} \mathbf{W} & \mathbf{K}_{21}^\top \\ \mathbf{K}_{21} & \mathbf{K}_{22} \end{bmatrix} \quad \text{and} \quad \mathbf{C} = \begin{bmatrix} \mathbf{W} \\ \mathbf{K}_{21} \end{bmatrix}.$$

The Nyström method uses  $\mathbf{W} \in \mathbb{R}^{l \times l}$ ,  $\mathbf{C} \in \mathbb{R}^{m \times l}$  and  $\mathbf{K} \in \mathbb{R}^{m \times m}$  to generate the approximation  $\tilde{\mathbf{K}} = \mathbf{C}\mathbf{W}^\dagger\mathbf{C}^\top \approx \mathbf{K}$ .

- (a) [5 points] Show that  $\mathbf{W}$  is symmetric positive semi-definite (SPSD) and that  $\|\mathbf{K} - \tilde{\mathbf{K}}\|_F = \|\mathbf{K}_{22} - \mathbf{K}_{21}\mathbf{W}^\dagger\mathbf{K}_{21}^\top\|_F$ , where  $\|\cdot\|_F$  is the Frobenius norm.

**Solution:** For the first part of question, note that  $\mathbf{W}$  is SPSP if  $\mathbf{x}^\top \mathbf{W} \mathbf{x} \geq 0$  for all  $\mathbf{x} \in \mathbb{R}^l$ . This condition is equivalent to  $\mathbf{y}^\top \mathbf{K} \mathbf{y} \geq 0$  for all  $\mathbf{y} \in \mathbb{R}^m$  where  $y_i = 0$  for  $l+1 \leq i \leq m$ . Since  $\mathbf{K}$  is SPSP by assumption, this latter condition holds. For the second part, we write  $\tilde{\mathbf{K}}$  in block form as

$$\tilde{\mathbf{K}} = \begin{bmatrix} \mathbf{W} \\ \mathbf{K}_{21} \end{bmatrix} \mathbf{W}^\dagger \begin{bmatrix} \mathbf{W} & \mathbf{K}_{21}^\top \end{bmatrix} = \begin{bmatrix} \mathbf{W} & \mathbf{K}_{21}^\top \\ \mathbf{K}_{21} & \mathbf{K}_{21}\mathbf{W}^\dagger\mathbf{K}_{21}^\top \end{bmatrix}.$$

Comparison with the block form of  $\mathbf{K}$  then immediately yields the desired result.

- (b) *[10 points]* Let  $\mathbf{K} = \mathbf{X}^\top \mathbf{X}$  for some  $\mathbf{X} \in \mathbb{R}^{N \times m}$ , and let  $\mathbf{X}' \in \mathbb{R}^{N \times l}$  be the first  $l$  columns of  $\mathbf{X}$ . Show that  $\tilde{\mathbf{K}} = \mathbf{X}^\top \mathbf{P}_{U_{\mathbf{X}'}} \mathbf{X}$ , where  $\mathbf{P}_{U_{\mathbf{X}'}}$  is the orthogonal projection onto the span of the left singular vectors of  $\mathbf{X}'$ .

Solution: Observe that  $\mathbf{C} = \mathbf{X}^\top \mathbf{X}'$  and  $\mathbf{W} = \mathbf{X}'^\top \mathbf{X}'$ . Thus,

$$\tilde{\mathbf{K}} = \mathbf{C} \mathbf{W}^\dagger \mathbf{C}^\top = \mathbf{X}^\top \mathbf{X}' (\mathbf{X}'^\top \mathbf{X}')^\dagger \mathbf{X}'^\top \mathbf{X} = \mathbf{X}^\top \mathbf{U}_{\mathbf{X}'} \mathbf{U}_{\mathbf{X}'}^\top \mathbf{X} = \mathbf{X}^\top \mathbf{P}_{U_{\mathbf{X}'}} \mathbf{X}.$$

(c) *[5 points]* Is  $\tilde{\mathbf{K}}$  SPSPD?

Solution: Yes. Using the expression for  $\tilde{\mathbf{K}}$  in (b) and the idempotency of orthogonal projection matrices, we can write  $\tilde{\mathbf{K}} = \mathbf{X}^\top \mathbf{P}_{U_{X'}} \mathbf{X} = \mathbf{A}^\top \mathbf{A}$ , where  $\mathbf{A} = \mathbf{P}_{U_{X'}} \mathbf{X}$ .

- (d) *[5 points]* If  $\text{rank}(\mathbf{K}) = \text{rank}(\mathbf{W}) = r \ll m$ , show that  $\tilde{\mathbf{K}} = \mathbf{K}$ . Note: this statement holds whenever  $\text{rank}(\mathbf{K}) = \text{rank}(\mathbf{W})$ , but is of interest mainly in the low-rank setting.

Solution: Since  $\mathbf{K} = \mathbf{X}^\top \mathbf{X}$ ,  $\text{rank}(\mathbf{K}) = \text{rank}(\mathbf{X}) = r$ . Similarly,  $\mathbf{W} = \mathbf{X}'^\top \mathbf{X}'$  implies  $\text{rank}(\mathbf{W}) = \text{rank}(\mathbf{X}') = r$ . The columns of  $\mathbf{X}'$  are columns of  $\mathbf{X}$  and they thus span the columns of  $\mathbf{X}$ . Hence,  $\mathbf{U}_{X'}$  is an orthonormal basis for  $\mathbf{X}$ , i.e.,  $\mathbf{I}_N - \mathbf{P}_{U_{X'}} \in \text{Null}(\mathbf{X})$ , and by part (b) of this exercise we have  $\mathbf{K} - \tilde{\mathbf{K}} = \mathbf{X}^\top (\mathbf{I}_N - \mathbf{P}_{U_{X'}}) \mathbf{X} = \mathbf{0}$ .

- (e) *[5 points]* If  $m = 20\text{M}$  and  $\mathbf{K}$  is a dense matrix, how much space is required to store  $\mathbf{K}$  if each entry is stored as a double? How much space is required by the Nyström method if  $l = 10\text{K}$ ?

Solution: Storage of  $\mathbf{K}$  requires roughly 3200 TB, i.e.,

$$(20 \times 10^6)^2 \text{ entries} \times 8 \text{ bytes/entry} \times \frac{1\text{TB}}{10^{12} \text{ bytes}} = 3200 \text{ TB}.$$

Storage of  $\mathbf{C}$  requires roughly 160 GB, i.e.,

$$(20 \times 10^6 \times 10^4) \text{ entries} \times 8 \text{ bytes/entry} \times \frac{1\text{GB}}{10^9 \text{ bytes}} = 1600 \text{ GB}.$$

Note that the computed numbers do not account for the symmetry of  $\mathbf{K}$  (doing so would change the storage requirements by less than a factor of two)