



10-605: Machine Learning with Large Datasets

Fall 2020

Recitation 4
25th September, 2020




Quick Review - SPSD matrices

A symmetric matrix $K \in \mathbb{R}^{n \times n}$ is said to be positive semidefinite iff all its eigenvalues are non-negative.

Furthermore, for any symmetric matrix K , the following statements are equivalent:

1. $z^T K z \geq 0$ for every vector z
2. There exists a matrix A such that $A^T A = K$
3. All the eigenvalues are non-negative



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.

Note: You can assume that \mathbf{W} is full-rank.

- 
- (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}' .



(c) *[5 points]* Is $\tilde{\mathbf{K}}$ symmetric positive semi-definite (SPSD)?

- 
- (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.



- (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 $l = 10\text{K}$?



Additional References

Linear Algebra Review:

1. 3Blue 1Brown - https://www.youtube.com/watch?v=fNk_zzaMoSs&list=PLZHQObOWTQDPD3MizzM2xVFitgF8hE_ab

Projection Matrices:

1. Piazza discussion post - <https://piazza.com/class/kcxot365ufc53w?cid=161>
2. https://ocw.mit.edu/courses/mathematics/18-06sc-linear-algebra-fall-2011/least-squares-determinants-and-eigenvalues/projections-onto-subspaces/MIT18_06SCF11_Ses2.2sum.pdf
3. <https://jekyll.math.byuh.edu/courses/m343/handouts/svdbasis.pdf>