# homework1

January 25, 2019

# 1 Homework 1 - Berkeley STAT 157

Handout 1/22/2017, due 1/29/2017 by 4pm in Git by committing to your repository. Please ensure that you add the TA Git account to your repository.

- 1. Write all code in the notebook.
- 2. Write all text in the notebook. You can use MathJax to insert math or generic Markdown to insert figures (it's unlikely you'll need the latter).
- 3. **Execute** the notebook and **save** the results.
- 4. To be safe, print the notebook as PDF and add it to the repository, too. Your repository should contain two files: homework1.ipynb and homework1.pdf.

The TA will return the corrected and annotated homework back to you via Git (please give rythei access to your repository).

In [1]: from mxnet import ndarray as nd

# 1.1 1. Speedtest for vectorization

Your goal is to measure the speed of linear algebra operations for different levels of vectorization. You need to use wait\_to\_read() on the output to ensure that the result is computed completely, since NDArray uses asynchronous computation. Please see http://beta.mxnet.io/api/ndarray/\_autogen/mxnet.ndarray.NDArray.wait\_to\_read.html for details.

- 1. Construct two matrices A and B with Gaussian random entries of size  $4096 \times 4096$ .
- 2. Compute C = AB using matrix-matrix operations and report the time.
- 3. Compute C = AB, treating A as a matrix but computing the result for each column of B one at a time. Report the time.
- 4. Compute C = AB, treating A and B as collections of vectors. Report the time.
- 5. Bonus question what changes if you execute this on a GPU?

In [ ]:

#### 1.2 2. Semidefinite Matrices

Assume that  $A \in \mathbb{R}^{m \times n}$  is an arbitrary matrix and that  $D \in \mathbb{R}^{n \times n}$  is a diagonal matrix with nonnegative entries.

- 1. Prove that  $B = ADA^{\top}$  is a positive semidefinite matrix.
- 2. When would it be useful to work with B and when is it better to use A and D?

### 1.3 3. MXNet on GPUs

- 1. Install GPU drivers (if needed)
- 2. Install MXNet on a GPU instance
- 3. Display !nvidia-smi
- 4. Create a  $2 \times 2$  matrix on the GPU and print it. See http://d2l.ai/chapter\_deep-learning-computation/use-gpu.html for details.

In []:

# 1.4 4. NDArray and NumPy

Your goal is to measure the speed penalty between MXNet Gluon and Python when converting data between both. We are going to do this as follows:

- 1. Create two Gaussian random matrices A, B of size  $4096 \times 4096$  in NDArray.
- 2. Compute a vector  $\mathbf{c} \in \mathbb{R}^{4096}$  where  $c_i = ||AB_{i\cdot}||^2$  where  $\mathbf{c}$  is a **NumPy** vector.

To see the difference in speed due to Python perform the following two experiments and measure the time:

- 1. Compute  $||AB_{i\cdot}||^2$  one at a time and assign its outcome to  $\mathbf{c}_i$  directly.
- 2. Use an intermediate storage vector **d** in NDArray for assignments and copy to NumPy at the end.

In []:

# 1.5 5. Memory efficient computation

We want to compute  $C \leftarrow A \cdot B + C$ , where A, B and C are all matrices. Implement this in the most memory efficient manner. Pay attention to the following two things:

- 1. Do not allocate new memory for the new value of *C*.
- 2. Do not allocate new memory for intermediate results if possible.

In []:

# 1.6 6. Broadcast Operations

In order to perform polynomial fitting we want to compute a design matrix A with

$$A_{ij} = x_i^j$$

Our goal is to implement this **without a single for loop** entirely using vectorization and broadcast. Here  $1 \le j \le 20$  and  $x = \{-10, -9.9, \dots 10\}$ . Implement code that generates such a matrix.

In []: