# CS583A: Coverage of the Final Exam

# March 10, 2019

# 1 Vector and matrix basics.

- Definitions of the  $\ell_p$  vector norms for p > 0.
- Definitions and properties of the  $\ell_p$ -norm balls.
- Matrix trace and matrix norms.
- Singular value decomposition (SVD) and truncated SVD.
- The BLAS and LAPACK libraries for matrix computation. (What are level 1, 2, and 3 BLAS?)

# 2 Scalar, vector, and matrix calculus.

- Derivative of a scalar w.r.t. a vector.
- Derivative of a vector w.r.t. a scalar.
- Derivative of a vector w.r.t. a vector.
- Derivative of a scalar w.r.t. a matrix.
- Chain rule.
- Always check the shape of a derivative!
- Subgradient and subdifferential.

# 3 Convex sets, convex functions, and convex optimization.

#### Convex sets.

- Definition of convex set.
- Typical examples of convex set and nonconvex set.

#### Convex functions.

- Definition of convex function.
- Definition of Hessian matrix.
- Definition of positive semi-definite.
- For convex function, the Hessian matrix is everywhere positive semi-definite.

## Convex and nonconvex optimization.

- Definition of convex optimization.
- Definitions of objective function, constrains, and feasible set.
- For convex optimization, local optimum is global optimum.
- For convex optimization, the first-order optimality condition  $(0 \in \partial f(\mathbf{w}^*))$  implies  $w^*$  is a global minimum.
- For nonconvex optimization, there are saddle points.
- Definition of saddle points.
- For high-dimensional nonconvex optimization, #saddle points >> #local minima >> #global minima.

# 4 Machine Learning Basics

#### The four ML tasks.

- Definitions of regression, classification, clustering, and dimensionality reduction.
- Difference between regression and classification.
- Supervised learning and unsupervised learning.

#### Classification.

- Binary classification and multi-class classification methods, e.g., logistic regression, support vector machine (SVM), softmax classifier, and k-nearest neighbor (KNN).
- Linear classifiers include logistic regression, SVM, and softmax classifier.
- Nonlinear classifiers include KNN, kernel SVM, neural networks.
- What classification method is most suitable if #classes is millions?
- Standard evaluation metrics: accuracy, classification error rate, top 1 classification error, top 5 classification error.
- Evaluation metrics for class-imbalanced problems: true positive, true negative, false positive, false negative, ROC curve, precision, and recall.

## Clustering.

- Clustering tasks are unsupervised learning.
- The k-means clustering method (a combinatorial optimization model).
- Lloyd's algorithm for approximately solving the k-means model.

## Dimensionality reduction.

- Unsupervised learning methods: PCA and autoencoder.
- Supervised learning method: linear discriminant analysis.

## Model capacity, overfitting, and underfitting.

- What controls model capacity? E.g., degree of polynomial in polynomial regression, number of layers and width of layers in neural networks, etc.
- What are overfitting and underfitting?
- How to alleviate overfitting and underfitting? More training samples, regularization, and data augmentation.

#### Hyper-parameters and cross-validation.

- Examples of hyper-parameters: degree of polynomials, regularization parameter, neural network structure, optimization algorithms. (The model itself is a hyper-parameter.)
- Training set, validation set, and test set.
- Never use the test for hyper-parameter tuning.

# 5 Convolutional Neural Networks

## Convolutional operations

- Definitions of patch, filter, matrix and tensor convolution, zero-padding, stride, etc.
- Calculate the output shape given the input shape, filter shape, stride, and zero-padding.

#### Convolutional neural networks

- Using Keras to implement convolutional layer, pooling layer, flatten layer, and dense layers.
- Being able to choose appropriate activation functions.
- Given the input shape, filter number, filter shape, stride, zero-padding, and pool size, infer the **output shape** and **number of parameters**.

- Tricks for allevating overfitting: regularization, data augmentation, and pretrain (with the bottom layers frozen).
- Other tricks for improving the test error: multi-task learning and ensemble method.
- What is dropout? What is the best place to insert a dropout layer?
- How to properly use pretrain? How to use fine-tuning?
- What is feature scaling? What is batch normalization?

# 6 Recurrent Neural Networks