# STAT430: Machine Learning for Financial Data

Fundamentals and workflows

# Four branches of machine learning

### Supervised learning

- Classification
- Regression
- Sequence generation: Given a picture, predict a caption describing it
- Syntax tree prediction: Given a sentence, predict its decomposition into a syntax tree
- Object detection: Given a picture, draw a bounding box around certain objects inside the picture
- Image segmentation: Given a picture, draw a pixel-level mask on a specific object

#### Unsupervised learning

- Dimensionality reduction and clustering
- Often a necessary step in better understanding a dataset before supervised-learning
- Self-supervised learning: supervised learning w/o human-annotated label, and labels are generated from data via some algorithms
- Reinforcement learning: an agent receives information about its environment and learns to choose actions that will maximize some reward

# Data preprocessing for neural networks

- · Vectorization: All inputs and targets in a neural network must be tensors of floating-point data
- · Value normalization
  - Take small value: most values should be in the 0–1 range
  - Be homogeneous: all features should take values in roughly the same range
  - R:scale()
- Missing values
  - Assign 0 for missing values if 0 is not a meaningful value
  - If missing in test set, but not in training set:
    - Copy some training samples several times, and drop some features from the training samples that are likely to be missing in the test set

# Feature engineering in the era of deep learning

- · Not as important as that for shallow learning, but can be beneficial:
  - Good features allow using fewer resources to solve problems more elegantly
  - Good features allow solving a problem with far less data

## Overfitting and regularization

- · Overfitting is always an issue in machine learning
  - The best solution is to get more training data
  - The 2nd best solution is **regularization**
- Regularization
  - Reduce learnable parameters / capacities
  - Weight regularization: adding to the loss function of the network a cost associated with weights
    - L1 regularization: proportional to the absolute value of the weights

```
- layer_dense(..., kernel_regularizer = regularizer_l1(0.001), ...)
```

- L2 regularization: proportional to the square of the value of the weights

```
- layer_dense(..., kernel_regularizer = regularizer_12(0.001), ...)
```

- layer\_dense(..., kernel\_regularizer = regularizer\_l1\_l2(l1 = 0.001, l2 = 0.001), ...)
- The cost is only added at training time, so the loss for this network will be much higher at training time than at test time
- Dropout

# Dropout

- · Applied to a layer, randomly setting to zero a number of output features of the layer during training
  - eg: for a given layer, the output is [0.2, 0.5, 1.3, 0.8, 1.1], and after dropout it becomes [0, 0.5, 1.3, 0, 1.1]
- Dropout rate: the fraction of the features that are zeroed out:  $(0.2 \sim 0.5)$
- Apply layer\_dropout(rate = 0.5) immediately before the layer
- · Try R

# Workflow of machine learning

- Defining the problem and assembling a dataset
- Choosing a measure of success
- Deciding on an evaluation protocol
  - Maintaining a hold-out validation set: when plenty of data is avaliable
  - Doing K-fold cross-validation: when too few samples for hold-out validation
  - Doing iterated K-fold validation: when little data is available
- Preparing data
- · Developing a model that does better than a baseline
  - two preliminary hypothesis:
    - outputs can be predicted by inputs
    - the available data is sufficiently informative to learn the relationships between inputs and outputs
  - Last-layer activation / Loss function / optimizers
- · Developing a model that overfits: add more and bigger layers, train for more epoches
- Regularizing your model and tuning your hyperparameters

# Last-layer activation and loss function

- Binary classification
  - sigmoid | binary\_crossentropy
- · Multiclass, single-label classification
  - softmax | categorical\_crossentropy
- · Multiclass, multilabel classification (non-exclusive labels for each observation)
  - sigmoid | binary\_crossentropy
- · Regression to arbitrary values
  - none | mse
- · Regression to values between 0 and 1
  - sigmoid | mse or binary\_crossentropy
- Back to Course Scheduler