## BT5152

AY2018/2019 Semester 1 Week 11

# W11: Support Vector Machine

## **Tutorial Logistics**

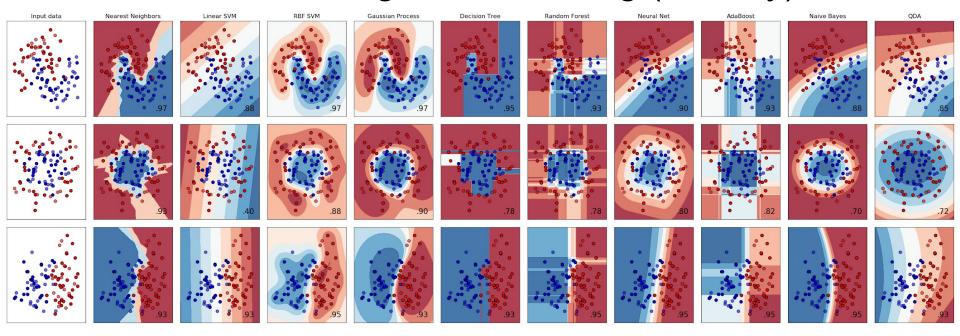
Swirl lesson on SVM is available on:

- https://github.com/weilu/BT5152
- https://github.com/kylase/BT5152

## Things we covering today

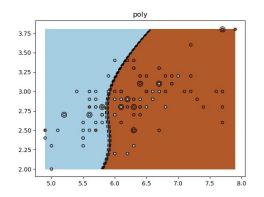
- 1. Brief talk on visualisation of ML
- 2. Training a linear SVM and visualise [Demo]
- 3. Training different SVM kernels and visualise them [10 to 20 mins]
- 4. Assignment 4 Reminders
- 5. Assignment 5

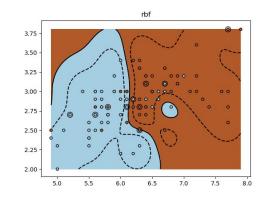
## What different ML algorithm is doing (visually)?

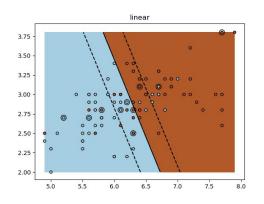


http://scikit-learn.org/stable/auto\_examples/classification/plot\_classifier\_comparison.html

### Different SVM Kernels

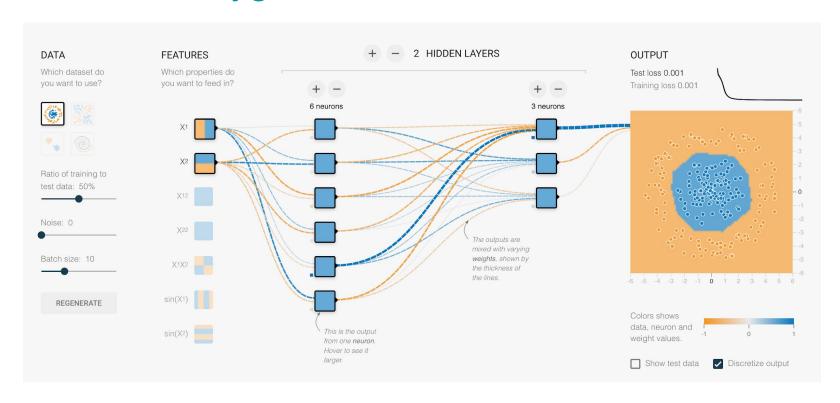




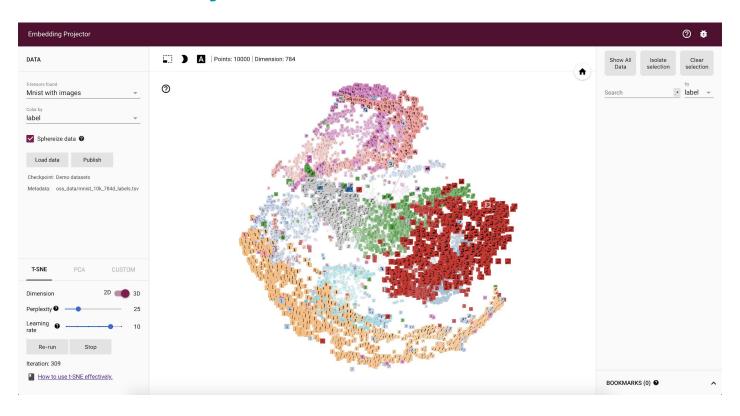


http://scikit-learn.org/stable/auto\_examples/exercises/plot\_iris\_exercise.html#sphx -glr-auto-examples-exercises-plot-iris-exercise-py

## Tensorflow Playground



## **Tensorboard Projector**



## Visualising Topic Model (my previous work)



## Sentiment Analysis (MovieLens)

#### **Naive Bayes**

Confusion Matrix and Statistics

#### Reference

Prediction neg pos neg 9001 2132 pos 3499 10368

Accuracy: 0.7748

95% CI: (0.7695, 0.7799)

No Information Rate : 0.5

P-Value [Acc > NIR] : < 2.2e-16

Kappa : 0.5495

Mcnemar's Test P-Value : < 2.2e-16

Sensitivity: 0.7201 Specificity: 0.8294

Pos Pred Value : 0.8085

Neg Pred Value : 0.7477

Prevalence: 0.5000
Detection Rate: 0.3600

Detection Prevalence : 0.4453 Balanced Accuracy : 0.7748

#### SVM (Linear Kernel)

Confusion Matrix and Statistics

#### Reference

Prediction neg pos neg 9984 2075 pos 2516 10425

Accuracy: 0.8164

95% CI: (0.8115, 0.8211)

No Information Rate : 0.5

P-Value [Acc > NIR] : < 2.2e-16

Kappa : 0.6327

Mcnemar's Test P-Value : 8.37e-11

Sensitivity: 0.7987

Specificity : 0.8340

Pos Pred Value : 0.8279 Neg Pred Value : 0.8056

Prevalence : 0.5000

Detection Rate: 0.3994
Detection Prevalence: 0.4824

Balanced Accuracy : 0.8164

## Training a SVM model using e1071

```
model <- svm(y \sim ., data=\underline{train}, kernel=<kernel>, cost=<cost of misclassification>, <kernel_parameters>)
```

#### Visualising the kernel and prediction

```
plot(model, train)
```

#### Training a linear kernel

```
model.linear <- svm(y \sim ., data=train, kernel='linear', cost=10)
```

## Assignment 4 Reminders

- Do not setwd()
- You can assume the data files are located in the data/ted-talk directory relative to where your code.R is
- If you are still stuck, please come to me after the class

## Assignment 5 Question 1

Using svm on A3 dataset (y  $\sim$  x4 + x8) with different kernels and visualise it.

- Polynomial
- RBF
- Sigmoid

## Assignment 5 Question 2

Home Credit Dataset (download from IVLE)

- Using XGBoost to predict TARGET with AUC as the optimisation metric to establish a baseline
- 2. Sampling to improve performance
  - Under-sampling
  - Over-sampling
  - Under/over-sampling together
  - SMOTE