- 1. Python (without Numpy) ⇒ [Module\_1, Module\_2(upto 12.37)]
  - a. Matrix Multiplication
  - b. Proportional Sampling
  - c. Replace numbers with #,

```
i. Ex 1: A = 234 Output: ###
ii. Ex 2: A = a2b3c4 Output: ###
```

- iii. Ex 3: A = abc Output: (empty string)
- iv. Ex 5: A = #2a\$#b%c%561# Output: ####
- d. Print name of students
  - i. who got top 5 ranks
  - ii. Who got least 5 ranks
  - iii. Who got marks in IQR
- e. Print 5 closest elements for a given point (p,q) based on the angle between (p,q) and (x,y)
- f. from the given set of hyperplanes find a hyperplane that will separate both red and green points
- g. Given two columns of data, the first column F will have 5 unique values and the second column S will have 3 unique values (0,1,2).

```
i. Find P(F_1|S==0), P(F_1|S==1), P(F_1|S==2)
```

ii. Find 
$$P(F_2|S==0)$$
,  $P(F_2|S==1)$ ,  $P(F_2|S==2)$ 

iii. Find 
$$P(F_3|S==0)$$
,  $P(F_3|S==1)$ ,  $P(F_3|S==2)$ 

iv. Find P(F 
$$4|S==0$$
), P(F  $4|S==1$ ), P(F  $4|S==2$ )

v. Find 
$$P(F 5|S==0)$$
,  $P(F 5|S==1)$ ,  $P(F 5|S==2)$ 

h. Filling the missing values in the specified format

```
i. _,_,value ex: _,_,40 \Rightarrow 10,10,10,10
ii. value, _,_,value ex:60,_,_,40 \Rightarrow 20,20,20,20,20
iii. Value, _,_, \Rightarrow 10,10,10,10
```

- i. Given two sentences S1, S2
  - i. Number of common words between S1, S2
  - ii. Words in S1 but not in S2
  - iii. Words in S2 but not in S1
- j. Given two columns of data Y, Y\_Score calculate the value of this function f(Y,Y\_score) = -1\*\frac{1}{n}\Sigma(Y Log10(Y\_score)+(1-Y)log10(1-Y\_score))
- 2. EDA  $\Rightarrow$  [Module 1, Module 2(upto 12.37)]
  - a. Get insights from data, given set of questions we need to answer them by doing a bit of analysis
  - b. Need two data sets of similar type one for reference, one for assignment
- 3. Implementing TFIDF vectorizer ⇒ [Module\_1, Module\_2, Module\_3(upto 18.14)]
  - a. Build a TFIDF Vectorizer, given the reference for countvectorizer
  - b. Implement min\_df and max\_feautres attributes

- Implement RandomSearchCV with k fold cross validation on KNN ⇒ [Module\_1, Module\_2, Module\_3(upto 19.31)]
  - a. For each hyper parameter select two disjoint set of indices and divide the data into train and test
  - b. Train model on train data and find the performance metric value on test data
  - c. Calculate the average performance metric score for each hyper parameter
- Compute Performance metrics without Sklearn ⇒ [Module\_1, Module\_2, Module\_3(upto 22.8)]

Given original and predicted values (without sklearn)

- a. #P >> #N
  - i. Calculate F1 Score
  - ii. Calculate AUC
  - iii. Calculate Accuracy
  - iv. Calculate confusion matrix
- b. #P << #N
  - i. Calculate F1 Score
  - ii. Calculate AUC
  - iii. Calculate Accuracy
  - iv. Calculate confusion matrix
- c. Find out the best threshold from given probability scores that will give the lowest score f(y,y^predict) = a\*fpr+b\*fnr, here y^pred = [0 if y\_score < y\_threshold else 1]
- d. Given y and y predicted(both are real-valued features)
  - i. Calculate MSE
  - ii. MAPE
  - iii. R Squared Error
- Apply Multinomial NB on Donors Choose Dataset ⇒ [Module\_1, Module\_2, Module\_3(upto 24.20)]
- 7. Implement SGD Classifier with Log Loss and L2 regularization Using SGD: without using sklearn ⇒ [Module\_1, Module\_2, Module\_3(upto 27.11)]

- 8. How each model behaves ⇒ [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4(upto 29.14)]
  - a. Given a highly imbalanced dataset (we will be incrementing imbalance in data adding few data points) observe how each model behaves
    - i. Draw hyperplane in Logistic regression
    - ii. Draw hyperplane in SVM
    - iii. Draw decision boundary in KNN
  - b. Given data with a set of outliers (we will be adding outliers in data incrementally)
    - i. How hyperplane changes in Logistic Regression
    - ii. How hyperplane changes in SVM
  - c. elliptical data with one or two outliers linear regression
  - d. Given 3d data points, such a way that var(3)>>(var2)>var(1)
    - How hyperplane changes before and after standardization of data in Logistic Regression
    - ii. How hyperplane changes before and after normalization of data in Logistic Regression
    - iii. How hyperplane changes before and after standardization of data in SVM
    - iv. How hyperplane changes before and after normalization of data in SVM
  - e. Create a dataset with features [X, X^2, 2\*X, Y, Z, X+Y] and perform perturbation test (iris.csv)
  - f. Visualization of the weight vector D1(X1, X2) with and without regularization D2(X1, X2, X3) with and without regularization
  - g. How the distribution of weights changes with increasing of lambda(L2) values for a dataset with 1000 features
- Apply Decision Trees on Donors Choose Dataset: [Module\_1, Module\_2, Module\_3, Module 4(upto 31.14)]
- 10. Application of Bootstrap samples in Random Forest: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4(upto 33.8)]
  - a. Choose any base model of your choice(either a Decision tree or Logistic regression). You can choose 30 to 40 base models based on your RAM.
  - b. Do both Row and column sampling to train each of the base learners.
  - c. Find the confidence interval on AUC based on the results of base learners
- 11. Apply GBDT/XGBOOST/LIGHT-GBM on Donors Choose Dataset: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4(upto 33.18)]
- 12. Clustering on Graph Dataset: [Module\_1, Module\_2, Module\_3, Module\_4, Module\_7(upto 45.9)]

- 13. Implement SGD algorithm to find the ratings that user is going to give to given movie. Provided reference notebook [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_7(upto 46.14) and case study 9]
- 14. Microsoft Malware detection Case Study assignment: [Module\_1, Module\_2, Module\_3, Module 4, Module 5, Module 6(refer case study 6)]
- 15. Facebook Assignment: [Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_6(refer case study 3)]
- 16. SQL Assignment: [Module\_1(upto 9.27)]
- 17. Implement a backpropagation on a given computation graph: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 50.14), <a href="https://www.youtube.com/watch?v=i94OvYb6noo">https://www.youtube.com/watch?v=i94OvYb6noo</a>, we will be providing reference videos and notebooks]
  - a. Reference will be given for a couple of computational graphs
- 18. Tensorflow Assignment, working with callbacks and vanishing gradient problem: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 50.14), we will be providing reference videos and notebooks]
- 19. Given an rvl-cdip dataset, classify the given document using transfer learning: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 53.18), we will be providing reference videos and notebooks]
  - a. Model 1: INPUT --> VGG-16 without Top layers(FC) --> Conv Layer --> Max pool Layer --> 2 FC layers --> Output Layer Train only new Conv block, FC layers, output layer. Don't train the VGG-16 network.
  - b. Model 2: INPUT --> VGG-16 without Top layers(FC) --> 2 Conv Layers identical to FC --> Output Layer Train only last 2 Conv layers identical to FC layers, 1 output layer. Don't train the VGG-16 network.
  - c. Model 3: 'INPUT --> VGG-16 without Top layers(FC) --> 2 Conv Layers identical to FC --> Output Layer' and train only Last 6 Layers of VGG-16 network, 2 Conv layers identical to FC layers, 1 output layer.
- 20. Classifying CIFAR-10 dataset images with DenseNet and work with optimization: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 53.18), we will be providing reference videos and notebooks]
  - a. Reference will be given for Dense-net architectures
- 21. Object detection YOLO pretrained model on image net dataset: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 53.18), we will be providing reference videos and notebooks]
- 22. CNN with text dataset: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 53.18), we will be providing reference videos and notebooks]
- 23. LSTM with Text and categorical data: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 54.10), we will be providing reference videos and notebooks]
  - a. Model 1: Glove embedding on text data, embedding layers on categorical features, dense layers for numerical features

- Model 2: Glove embedding on text data(consider words with TF IDF values within IQR), embedding layers on categorical features, dense layers for numerical features
- c. Model 3: Glove embedding on text data  $\rightarrow$  LSTM, one hot encode the categorical and merge all of them  $\rightarrow$  CNN1D. Merge both the outputs of LSTM and CNN1D
- 24. LSTM with Time series data: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 54.10), we will be providing reference videos and notebooks]
- 25. Encode-decoder Architecture for text abstraction, seq-seq: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 54.10), we will be providing reference videos and notebooks]
- 26. Siamese Architecture for Q and A selection: [We will be updating assignments soon, Module\_1, Module\_2, Module\_3, Module\_4, Module\_5, Module\_8(upto 54.10), we will be providing reference videos and notebooks]
- 27. Personal Case study -1: ML/RS
- 28. Personal Case study -2: DL
- 29. Blog on Personal Case study
- 30. Blog on given concept