

# OTT REVENUE PREDICTION MODEL (NETFLIX)

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# INTRODUCTION

- Over The Top platform
- Challenges
  - Multiple OTT platforms
  - High resolution video/audio
  - Content selection
  - Viewers retention
  - Revenue model
- OTT Revenue Models
  - SVOD (Subscription Video On Demand)
  - AVOD (Advertising Video On Demand)
  - Hybrid
- Problem Statement

## *OTT Revenue Prediction model*

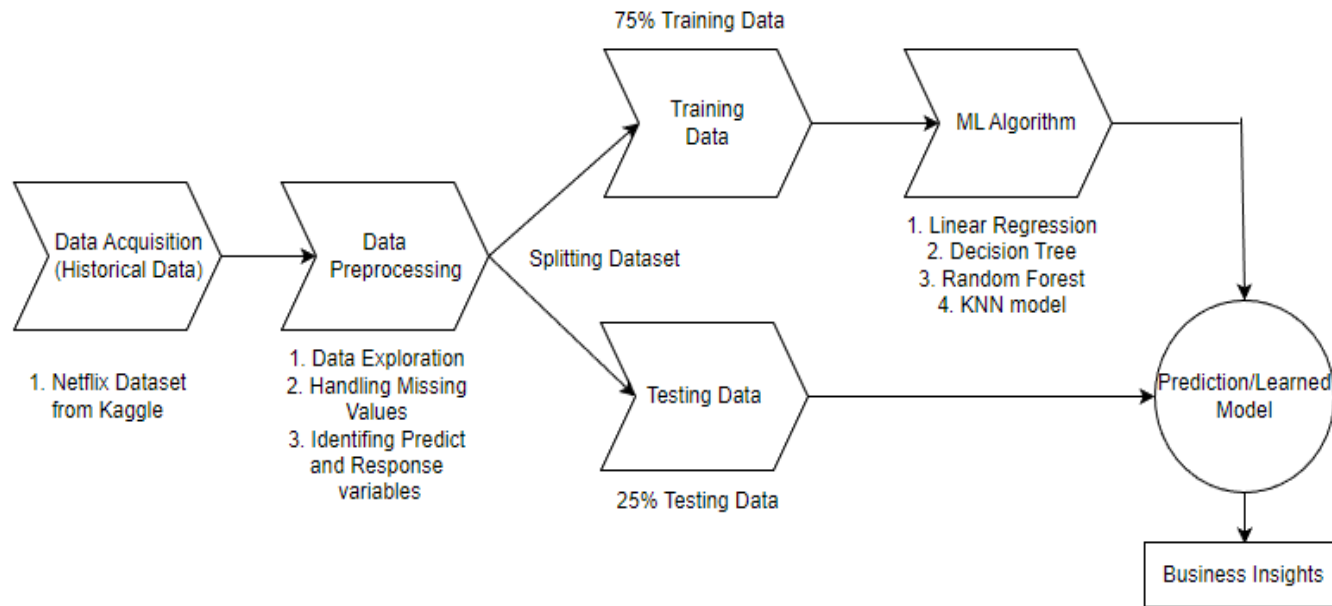
How much and how quickly a company can intend to grow?

How likely to plan for investments?

How quickly and what are the actions to be taken to improve business if there is a fall in revenue?

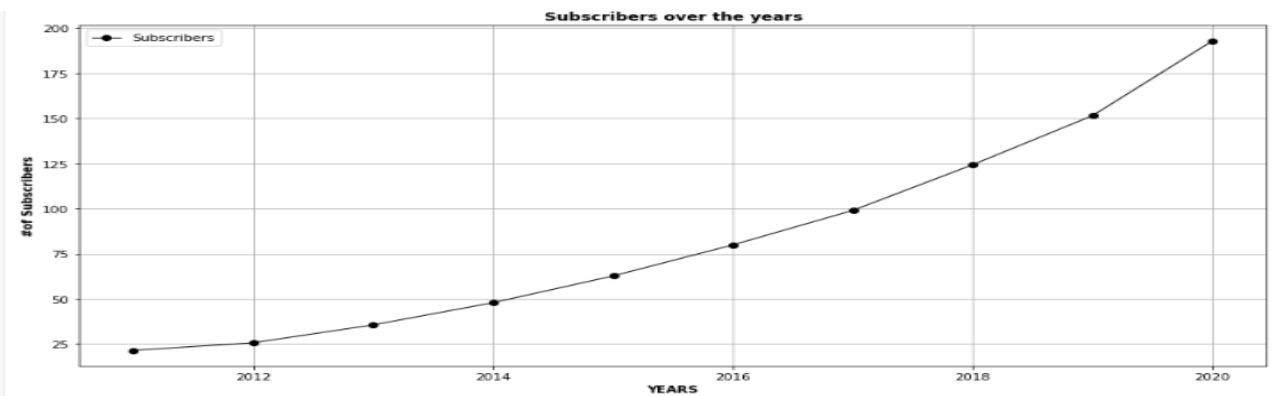
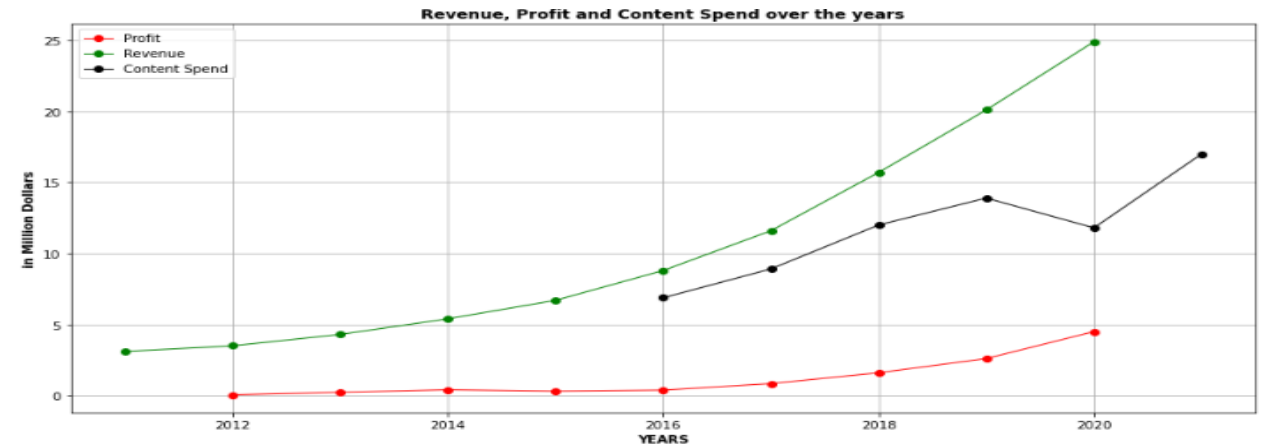
# PREDICTION PROCESS

- Predict the revenue of Netflix OTT platform using different models: Linear Regression, Decision Tree, Random Forest and KNN models



# DATA PRE-PROCESSING AND VISUALIZATION

- Understanding the data
  - > Dataset Link: [Kaggle-Netflix Revenue and Users](#).
  - > Independent variable X: Subscribers/Year/Content Spend/ in million dollars/Profit in million dollars
  - > Dependent variable Y: Overall revenue generated in million dollars
- Import required libraries
  - Pandas, NumPy, Matplotlib, Seaborn, Sci-Kit
- Import dataset and explore data
  - Data object creation and `read_csv()`
- Identifying and handling missing values
  - Impute median
- Extract dependent and independent variable
  - Relationship using `pairplot()` / `Implot()` functions
- Split the dataset
  - Training: Testing [75:25]



# REGRESSION MODELS

## 1. Linear Regression Model:

- ➔ LinearRegression () is imported from sci-kit library
- ➔ Feature selection is done by checking the MSE of each feature against revenue. The one with lower MSE is chosen for prediction (Graph 1).
- ➔ Training and testing scores are computed along with R2 score, MSE(Mean Square Error) and RMSE (Root Mean Square Error) which help evaluate the linear regression model

## 2. Decision Tree Regression Model:

- ➔ DecisionTreeRegressor () function is imported from sci-kit library

## 3. Random Forest Regression Model

- ➔ RandomForestRegressor () function is imported from sci-kit library

## 4. K- Nearest Neighbors Regression Model

- ➔ KNeighborsRegressor () function is imported from sci-kit library and chose nearest neighbors' parameter to be 2

For all the above models,

- ➔ Number of subscribers is taken as an independent variable to predict Revenue (target/dependent variable)
- ➔ Model is evaluated by computing training and testing scores, R2 score, MSE and RMSE

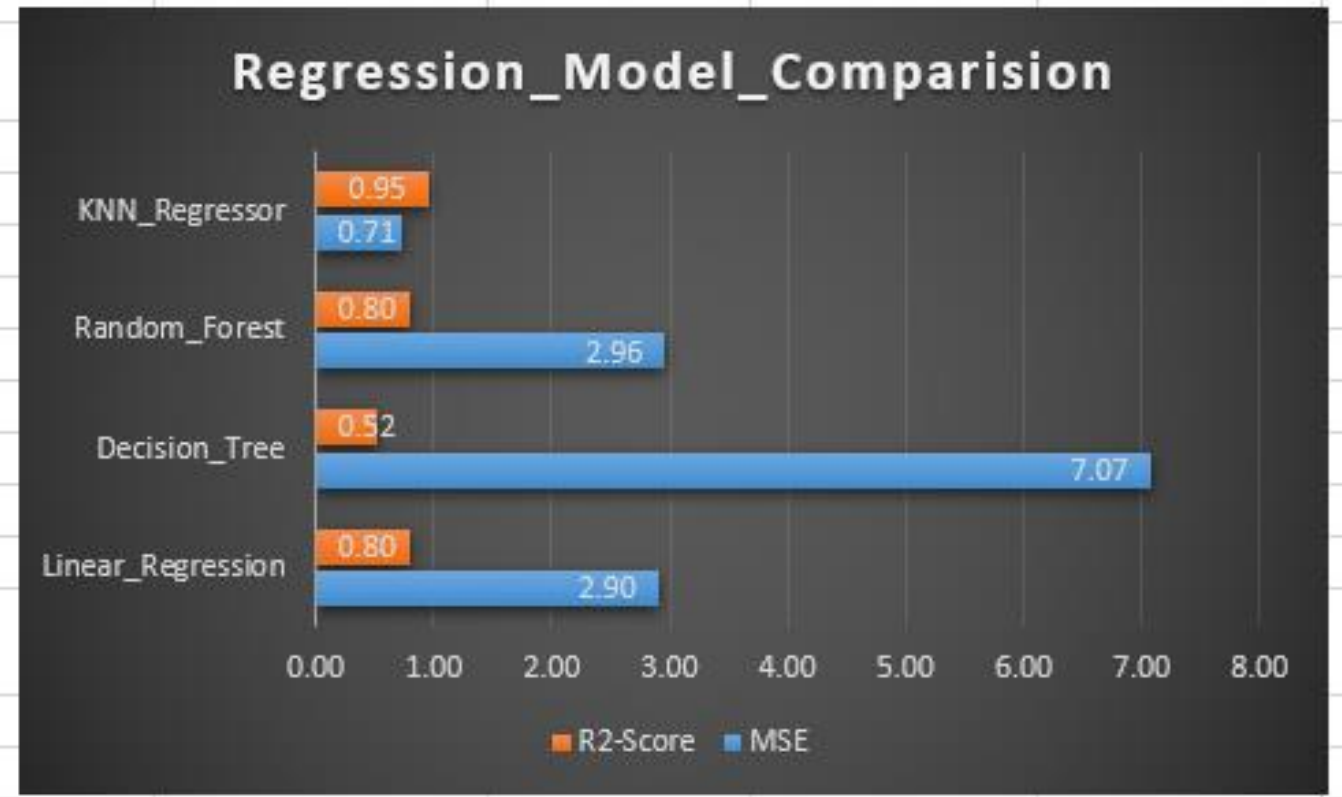
# EVALUATION: LINEAR REGRESSION

Linear_Regression_Model	MSE	Training Score	Testing Score
Year	14.92	0.41	0.66
Number of Subscribers	2.90	0.92	0.80
Profit	2.93	0.91	0.92
Content Spend	56.95	0.02	-0.19
X=(year, Number of subscribers, Profit, Content Spend)	105.93	0.95	0.43



# MODEL COMPARISON

#of_Subscribers	Linear_Regression	Decision_Tree	Random_Forest	KNN_Regressor
MSE	2.90	7.07	2.96	0.71
R2-Score	0.80	0.52	0.80	0.95



# CONCLUSION

- From the OTT point of view, these prediction models help in making business critical decisions and provide insights to identify trends in market.
- For a given set of data used in the project, Linear Regression model suits best
- Scope of fine tuning the hyper parameters in other regression models, we might see better predictions and lower errors.
- Deep Neural Networks could help make result more accurate.



## REFERENCES

[1]<https://www.kaggle.com/datasets/azminetoushikwasi/ott-video-streaming-platforms-revenue-and-users>

[2]<https://selectra.in/blog/how-ott-earn>

[3]<https://scikit-learn.org/>

[4]<https://pandas.pydata.org/>

[5]<https://matplotlib.org/>

[6]<https://numpy.org/>

[7]<https://seaborn.pydata.org/>

Any Question??

Thank You!!