



IU 4.6 - Machine Learning

Mini Project

Content

180 min

Where Are We in the Journey & Learning Objectives

5 min

Agenda for Today

- Mini Project: Introduction to problem statement
 - Q&A
 - Group activity to solve the project
-

30 min

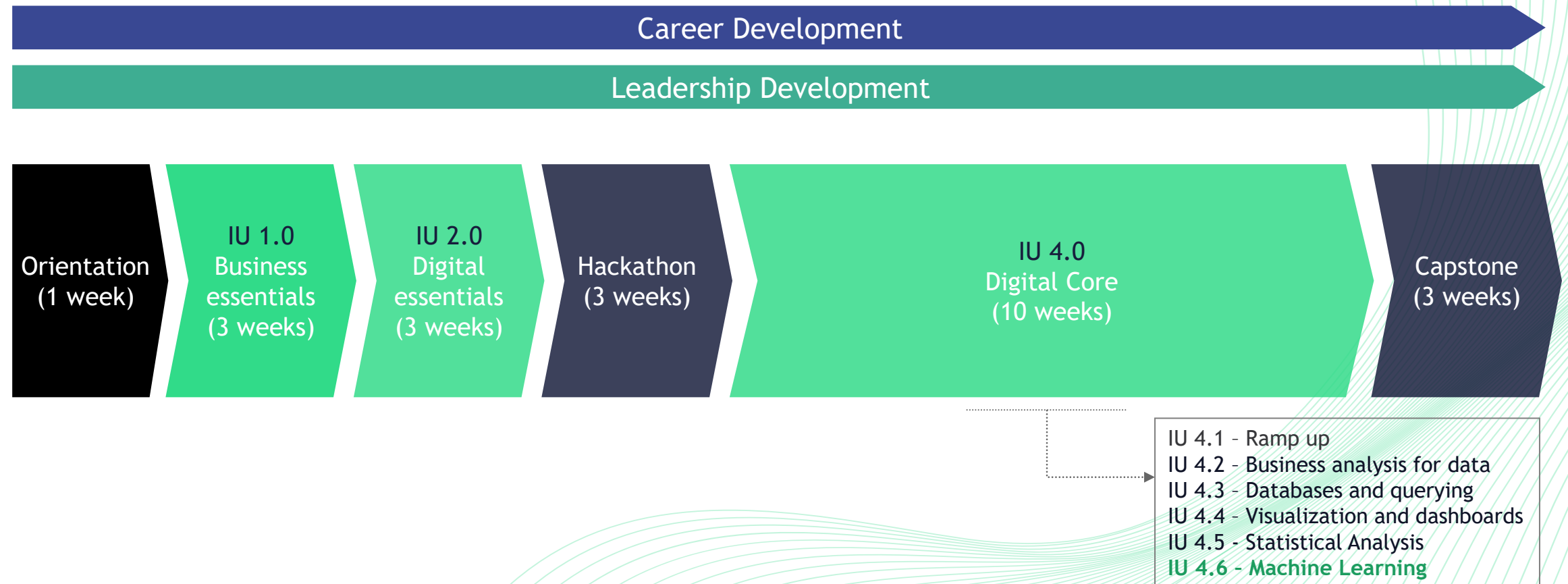
20 min

120 min

Wrap up

5 min

Where are we in the learning journey?



Learning objectives

- Upon successful completion of the mini project, individuals will be able to build and evaluate machine learning models for a classification based problem
- Will be able to apply various model evaluation techniques from technical and business perspective



01

Mini Project: Introduction to the problem statement

Major SEA Telecom Provider

Providing telecom services to prepaid and postpaid customer segments combined with variety of product offers and plans



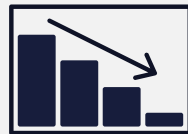


Business Problem

Company's postpaid business of voice only plans is struggling to maintain its strong foothold in local market because of,



High churn rate amongst customers leading to a revenue decline of ~500k USD every month



Decline in overall customer base (high churn rate combined with low acquisition rate), leading to a decline in total market share

Project objectives

Business Objective: Reduce monthly customer churn by identifying high risk customers well in advance

Hypothesis

- Company CEO believes that existing models can predict churners precisely, but it's too late to take any retention actions, as customers usage have significantly declined by then

Analytics Objective

1. Build a classification model to predict churners one month in advance
2. Identify key churn drivers



What is provided?

Data provided for 50k customers who are currently availing voice only postpaid plans from the telecom provider

Data Source	Data fields available
Customer profile	Months in service, Unique subscribers, Active subs, Service area, handset, model, Age HH1, Age HH2, Marital status, Occupation, Home ownership, Has Credit card, Owns motorcycle, Credit rating and changes over time, etc.
Customer Revenue	Monthly revenue, recurring revenue from prior month, Percentage change in revenue
Customer Usage and Activity	Monthly minutes, percent change in minutes, Overage minutes, Inbound & outbound calls, Peak & off-peak calls, dropped/blocked/unanswered calls, roaming calls, etc.
Customer Interaction	Customer care calls, Responds to mail offers, Retention calls, Retention offer accepted, referrals made, etc.
Churn	Churn/ Non-churn flag for each customer

Data Dictionary

Columns	Description	Columns	Description	Columns	Description
CustomerID	Unique Customer ID	CallForwardingCalls	Minutes in call forwarding	NonUSTravel	Flag indicating whether the customer has travelled outside of US
MonthlyRevenue	Monthly revenue in USD	CallWaitingCalls	Minutes spend on hold during call	OwnsComputer	whether customer Owns computers
MonthlyMinutes	Monthly minutes	MonthsInService	Total months in service	HasCreditCard	Whether customer owns a credit card
TotalRecurringCharge	Recurring charges in the past month	UniqueSubs	total number of unique sim card subscriptions (includes inactive connection)	RetentionCalls	Whether customer responded to retention calls
DirectorAssistedCalls	Automated calll (directory assisted calls)	ActiveSubs	total # of active subscriptions	RetentionOffersAccepted	Whether customer accepted retention offers
OverageMinutes	Extra minutes above the postpaid allocation	ServiceArea	Service area of the subscription	NewCellphoneUser	New cellphone user flag
RoamingCalls	Minutes on calls while roaming	Handsets	Total # of handsets	NotNewCellphoneUser	Total not new cellphone user flag
PercChangeMinutes	percentage change in minutes from previous month	HandsetModels	total # of unique hadnset model	ReferralsMadeBySubscriber	Total referrals made by subscriber
PercChangeRevenues	percentage change in revenue from previous month	CurrentEquipmentDays	Number of days since the activation of current equipment	IncomeGroup	Income group
DroppedCalls	Dropped calls in minutes	AgeHH1	Primary holder	OwnsMotorcycle	Owns motorcycle flag
BlockedCalls	Blocked calls in minutes	AgeHH2	Secondary holder	AdjustmentsToCreditRating	Number of time the credit rating has ranged in past 1 year
UnansweredCalls	Unanswered calls in minutes	ChildrenInHH	Flag indicating children in household	HandsetPrice	Price of the handset in USD
CustomerCareCalls	Customer care call duration in minutes	HandsetRefurbished	Handset refurbished flag (returned to company and then they sell it to different customer)	MadeCallToRetentionTeam	Flag indicating whether customer made call to retention team
ThreewayCalls	Minutes spend on Conference calls	HandsetWebCapable	Internet connectivity	CreditRating	Credit rating of cthe customer
ReceivedCalls	Total received calls in minutes	TruckOwner	Flag indicating whether the Customer owns a truck	PrizmCode	Area group of customer home location
OutboundCalls	Marketing calls received from customer service in minutes	RVOwner	Customer owns RV or not	Occupation	Type of occupation
InboundCalls	total duration in minutes of calls made to customer service	Homeownership	Home owned by customer	MaritalStatus	Marital status
PeakCallsInOut	Incoming/outgoing calls during peak time	BuysViaMailOrder	Whether customer has bought anything via clicking an option on email		
OffPeakCallsInOut	Incoming/outgoing calls during off peak time	RespondsToMailOffers	Flag indicating whether customer responds to mail offers		
DroppedBlockedCalls	Summation of dropped and blocked	OptOutMailings	Whether the customer has opted out of mailing		

Group activity | Assignment



Exercise Work plan

Estimated time
to complete: 5-7 hours

1. Detect and resolve problems in the data (Missing value, Outliers, Unexpected value, etc.) *[30 - 45 mins] [Marks: 10]*
 - i. How many customers had zero monthly revenue?
 - ii. How many columns have missing values percentage > 5%?
 - iii. For columns, "UniqueSubs" and "DirectorAssistedCalls" remove outliers, if any
2. Perform exploratory analysis to analyze customer churn *[30 - 45 mins] [Marks: 15]*
 - i. Does customers with high overage minutes also have high revenue?
 - ii. Does high number of active subscribers lead to low monthly revenue?
 - iii. Does credit rating have an impact in churn rate?
3. Create additional features to help predict churn *[20 - 40 mins] [Marks: 15]*
 - i. Percent of current active subs over total subs
 - ii. Percent of recurrent charge to monthly charge
 - iii. Percent of overage minutes over total monthly minutes
4. Build classification model to predict customer churn *[120 - 180 mins] [Marks: 50]*
 - i. Build a simple logistic regression model to predict churn and evaluate model accuracy on test data set
 - ii. Build Random Forest classifier to compare model accuracy over the logistic regression model
 - iii. Identify most important features impacting churn
(Model evaluation metrics to be used: GINI, AUC, Precision and Recall)
5. Use the hold out data provided to predict churners using the best model identified in step 4 *[45 - 60 mins] [Marks: 10]*
6. **Bonus Question:** Calculate lift chart and total monthly revenue saved by targeting top 10-20% of the customers using your best predictive model *[60 - 90 mins]*

What are we giving you?

Q1.2 How many columns has missing values percentage > 5%

```
In [9]: #Calculate for each column % of missing value in the data
#How many columns has missing values percentage > 5%
#What strategy should be used for imputation?
```

Result:

Type your answer here for how would you impute the missing values (if any)

Q1.3 For columns, "UniqueSubs" and "DirectorAssistedCalls" remove outliers, if any

```
In [15]: #plot box plot using pandas for columns "UniqueSubs" and "DirectorAssistedCalls"
cols=["UniqueSubs", "DirectorAssistedCalls"]
cust_df.boxplot(column=cols)
```

```
Out[15]: <AxesSubplot:>
```



Initial data processing for model building exercise

```
#Train - test split to train and test model accuracy
from sklearn.model_selection import train_test_split

#Define columns to be included in X and y
# X = Independent variables
# Y = Dependent variable (churn flag)

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3,

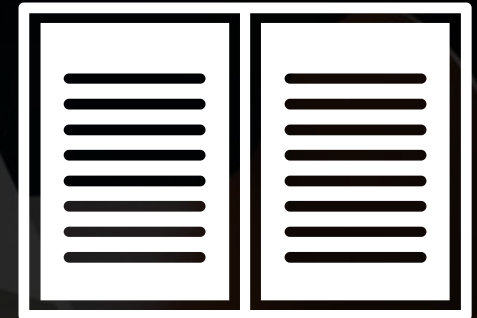
#Feature scaling for all continuous variable
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
```

Starter code provided in
Jupyter notebook



Guidelines on submitting mini project solution

- 1) Work during the live class with your respective groups to brainstorm ideas and solve analytics objectives based on the work plan provided; Divide tasks within the group where possible
- 2) **Submission guidelines** - All team members must submit their solution via LMS using the Jupyter notebook template provided; One member from the group to submit the solution via Microsoft Teams as well, specifying the group number
- 3) Final solution submission must be done during the final work and submission session
- 4) Grading will be done across - (i) Submission: 40% (ii) Concepts applied: 30% (iii) Insights/recommendation: 30%
- 5) Digital badges will be provided for top two teams based on grades achieved

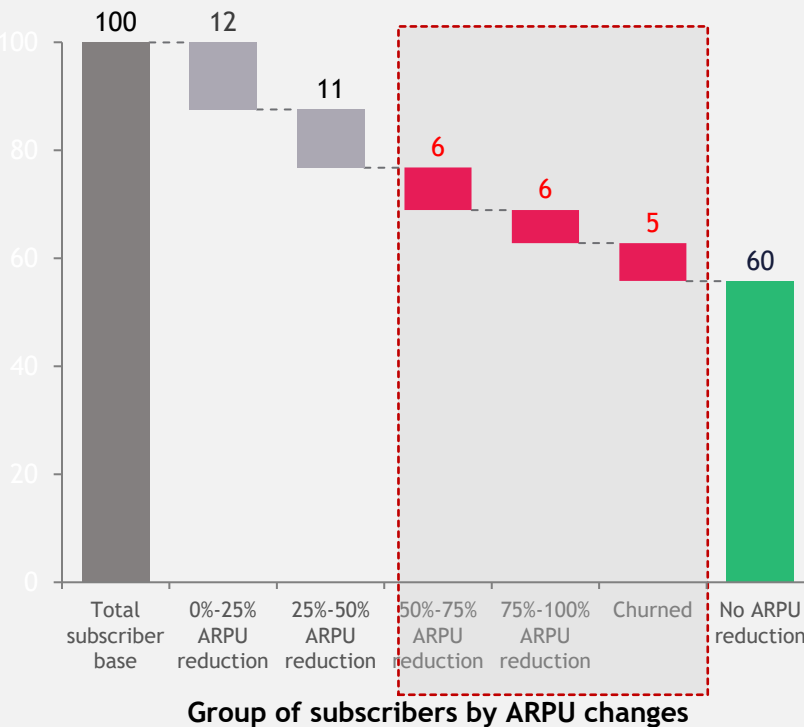


Recap Customer Churn - Typical business scenarios in Telco industry



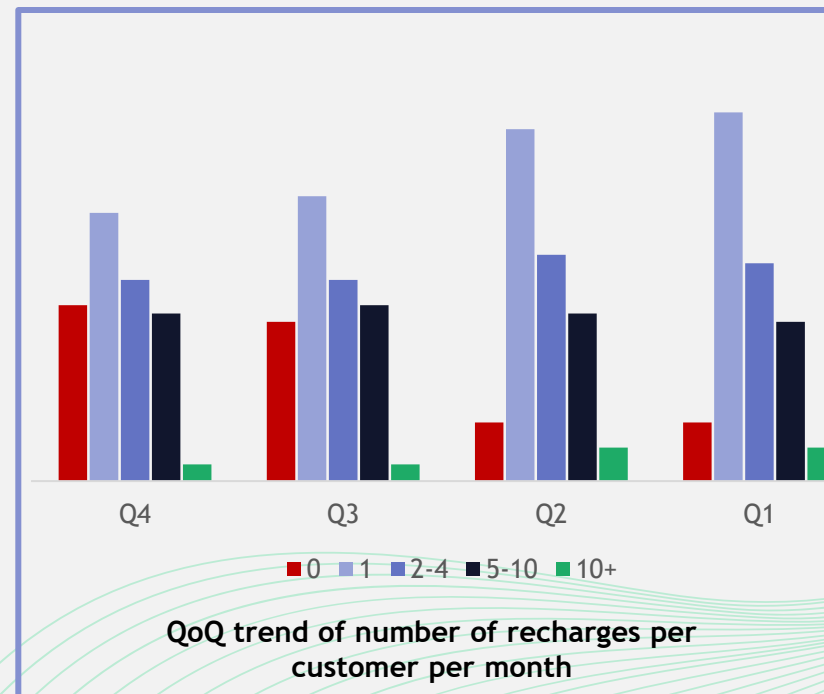
Steady decline in ARPU

About **X%** customers exhibit more than 50% drop in monthly ARPU for 3 months straight



Low Customer Engagement

It was observed that the average number of recharges per customer had gone down by **Y%** in last 6 months



Lack of Retention Strategy

The retention offers sent out to customers are driven by a basic rule-based framework

Key drivers for existing retention offer framework

- Last Recharge Amount
- Current Plan
- Best available discount

Retention offers sent as a part of the same campaigns as up-sell/growth

Recap: Machine Learning topics that you will need to use

1

Exploratory data analysis

- Missing value identification and treatment, outlier detection, etc.
- Univariate analysis - histograms to check distribution, box/violin plot, summary stats such as mean, median, mode, etc.
- Correlation analysis, scatter plots
- Time analysis to monitor trend and seasonality

2

Feature engineering

- Target variable creation (dependent variable) - align with analytics objective
- New feature creation based on business and analytics objectives
- Trend variables such as moving average, standard deviation over time, etc.

3

ML model build

- Model type - classification vs regression
- Algorithm type - logistic regression, linear regression, Random forest, etc.
- Train and test data split (e.g., 70:30 split)
- Hyperparameter tuning for model optimization
- Feature importance

4

Model evaluation

- Model evaluation metrics
- Regression - R-Square, Mean Absolute Error (MAE), Mean Square Error (MSE), etc.
- Classification - Confusion metrics, precision, recall, GINI, AUC, etc.
- Lift and Gain charts
- Hold out data set for model validation

20 Mins

02

Questions?

120 Mins

02

Working session - Group activity

