



Telecom Churn Case Study Doubts Session

Course : Data Science
Lecture On : Assignment
Instructor : Sumit Shukla

What we will cover in this session?

- 1 Case Study Walkthrough
- 2 QnA

Assignment

customer usage data

NO ready made target column provided.

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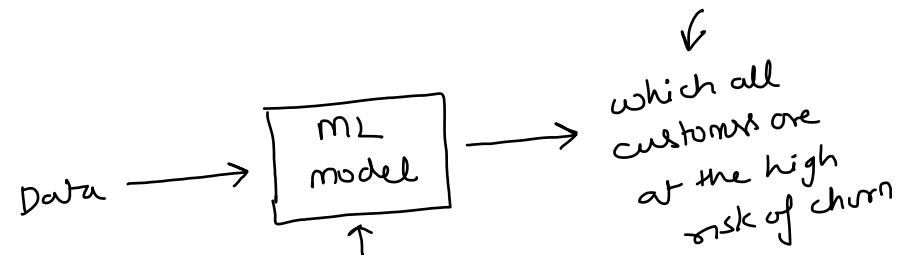
	6			7			8			9		
	x	y	z	x	y	z	x	y	z	x	y	z
c ₁												
c ₂												
...												

Problem Statement

6 } Good phase ← customers happy with service
7 }
8 → Action phase ← started facing some issues
1 → Churn phase ←

In the telecom industry, customers are able to choose from multiple service providers and actively switch from one operator to another. In this highly competitive market, the telecommunications industry experiences an average of 15-25% annual churn rate. Given the fact that it costs 5-10 times more to acquire a new customer than to retain an existing one, customer retention has now become even more important than customer acquisition.

Predict which customers are at high risk of churn.



Assignment

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Data
cleaning

What you need to do?

1. Handling Missing data.
 - a. Impute with zero when you are very sure that a missing is a zero.
 - b. For categorical, what to do?
 - c. Remove those with high missing percentage.

<https://www.kaggle.com/athi94/investigating-imputation-methods>

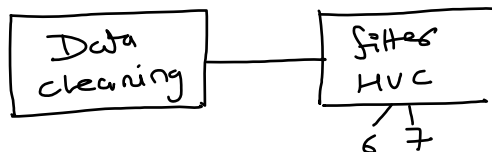
	Date of rech	amount of rech	# of rech	actual missing (mean/median)
C ₁	01/01/2011	55	NA	
C ₂	NA	NA	NA	

○ (NO recharge was made)
meaningful missing

- i) find and impute those rows and columns where a missing can be imputed with zero
- ii) find out percentage of missing data for various columns and drop those columns with high %age of missing
- iii) Impute columns with less percentage of missing

mean 'median' mode

Assignment



Filter high-value customers(HVC)

↳ top 30%

80/20 Pareto rule : 80% Revenue → top 20% customers

Good Phase

1. Calculate average recharge done by customer in June and July(total_rech_amt)
2. Look at the 70th percentile recharge amount
3. Retain only those customers who have recharged their mobiles with more than or equal to 70th percentile amount

total rech-amt 6	total rech-amt 7	avg-rech amt-6-7
a	b	$(a+b)/2$

fine
 $\left\{ \begin{array}{l} 30.1k \\ 29.9k \text{ rows} \\ 29.8k \end{array} \right.$

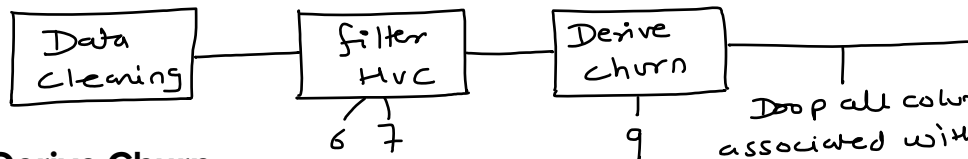
HVC ← $\left\{ \begin{array}{l} \text{avg-rech} \\ \text{amt-6-7} \end{array} \right. \geq 70^{\text{th}} \text{ percentile HVC}$

i) Create new column avg-rech-amt-6-7 which is the avg of total-rech-amt-6 & total-rech-amt-7

ii) find the 70th percentile value of for avg-rech-amt-6-7

iii) If avg-rech-amt-6-7 for a customer greater than or equal to 70th percentile value

Assignment



Derive Churn

9th Month is our Churn Phase. Usage-based churn

1. Calculate total incoming and outgoing minutes of usage
2. Calculate 2g and 3g data consumption
3. Create churn variable: those who have not used either calls or internet in the month of September are customers who have churned
4. Check Churn percentage.
5. Delete columns that belong to the churn month

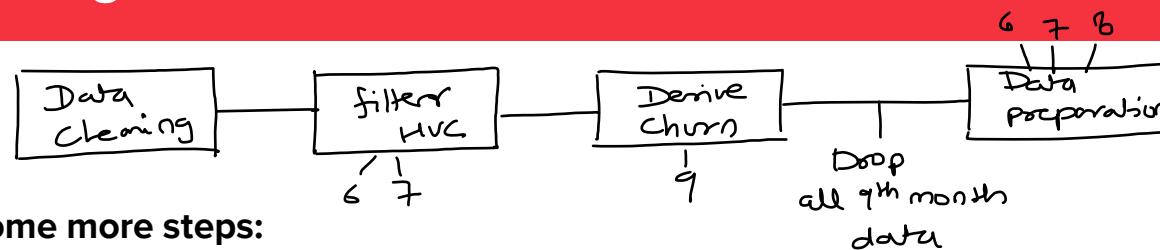
If for month 9

$$\left\{ \begin{array}{l} \text{total-incoming-min} == 0 \\ \text{total-outgoing-min} == 0 \\ \text{total-2g-data} == 0 \\ \text{total-3g-data} == 0 \end{array} \right.$$

Churn $\left\{ \begin{array}{l} \text{usage based churn} \rightarrow \text{customer is not at all active for any service} \\ \text{Revenue based churn} \rightarrow \text{using incoming \& SMS service but not making any revch / generating any revenue for the company} \end{array} \right.$

Assignment

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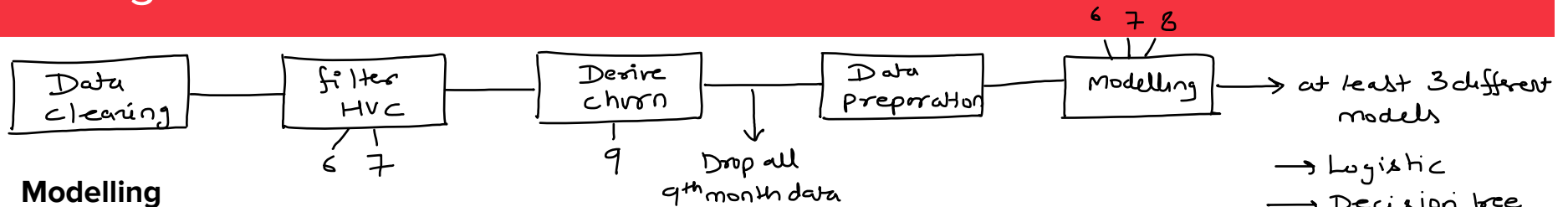


Some more steps:

1. Derived variable. at least 3 additional
2. EDA
3. Outlier treatment
4. Split train-test
5. Scaling

Assignment

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Modelling

1. Try out various models and select the best among them.
2. You need to handle the imbalance class for all the models.

<https://towardsdatascience.com/methods-for-dealing-with-imbalanced-data-5b761be45a18>

within
jupyter
notebook
markdown
comment

using the logistic
Regression model
add recommendations
for the company
so that they may
avoid churn

- Logistic
- Decision tree
- Random forest
- for all models handle class imbalance
- Hyperparameter tuning
- model evaluation
- Choose the best model according to the sensitivity

