TELECOM CHURN CASE STUDY

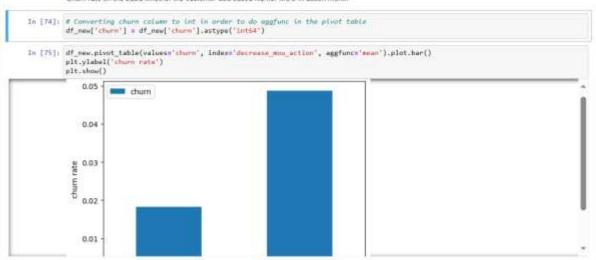
PRESENTED BY SAKSHI BHASIN RUCHI GUPTA RUCHIRA DOLAI

DS C54
BUSINESS ANALYTICS

EDA Analysis

Step 3.1 : Univariate analysis

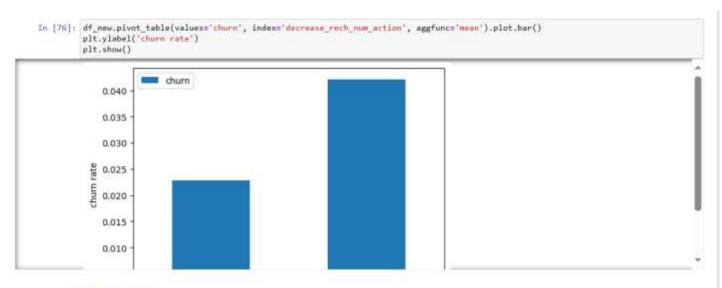
Churn rate on the basis whether the customer decreased his/her MOU in action month



Observations:

We can observe that the churn rate is more for the customers, whose minutes of usage(mou) decreased in the action phase compared to that when the customer is in the good phase.

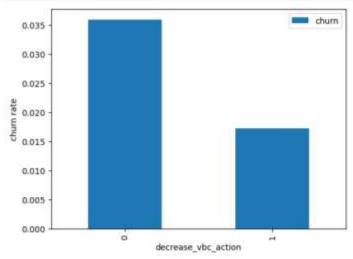
Churn rate on the basis whether the customer decreased her/his number of recharge in action month



We can observe that the churn rate is more for the customers, whose number of recharge in the action phase is lesser than the number in good phase.

Churn rate on the basis whether the customer decreased her/his amount of recharge in action month





We can observe that the churn rate is more for the customers, whose volume based cost in action month is increased which means that the customers do not do the monthly recharge more when they are in the action phase.

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In [79]: # Creating them dataprisms
charm, df - df need[df_nee] charm'] -- 1]

# (restring not charm dataprisms
non_charm_df - df_need[df_need] charm'] -- 0]

In [88]: # pistribution piot
as = sin_distplot(pen_df] avg_arpu_action'], label-'sharm', hist-false)
as = sin_distplot(pen_dium_df] avg_arpu_action'], label-'not charm', hist-false)
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as_sin_distribution avg_arpu_action', label-'not charm', hist-false
```

0.00000

We can observe that the Average revenue per user (ARPU) for the channed customers is mostly densed on the 0 to 900. The higher ARPU customers are less likely to be channed

3000

4000

ARPU for the not churned customers is mostly densed on the 0 to 1000.

1000

2000

Action phase ARPU

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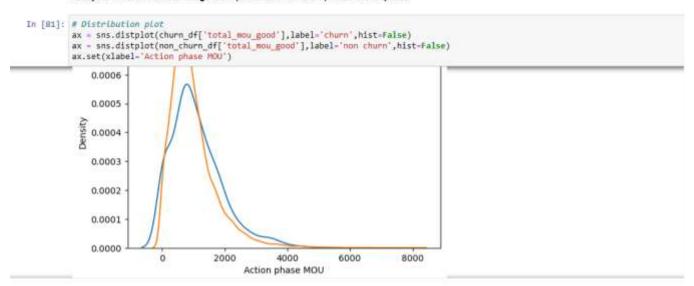
ARPU for the not churned customers is mostly densed on the 0 to 1000.

1000

2000

Action phase ARPU

Analysis of the minutes of usage MOU (churn and not churn) in the action phase



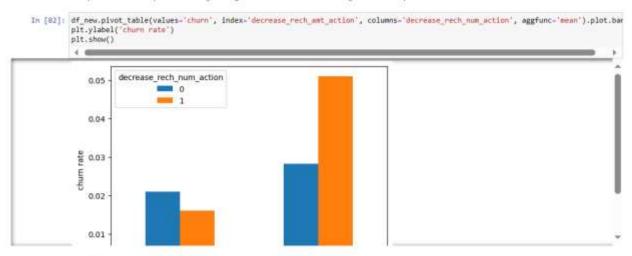
Observations:

We can observe that the Minutes of usage(MOU) of the churn customers is mostly populated on the 0 to 2500 range. Higher the MOU, lesser the churn probability.

Step 3.2: Bivariate analysis

Step 3.2 : Bivariate analysis

Analysis of chum rate by the decreasing recharge amount and number of recharge in the action phase

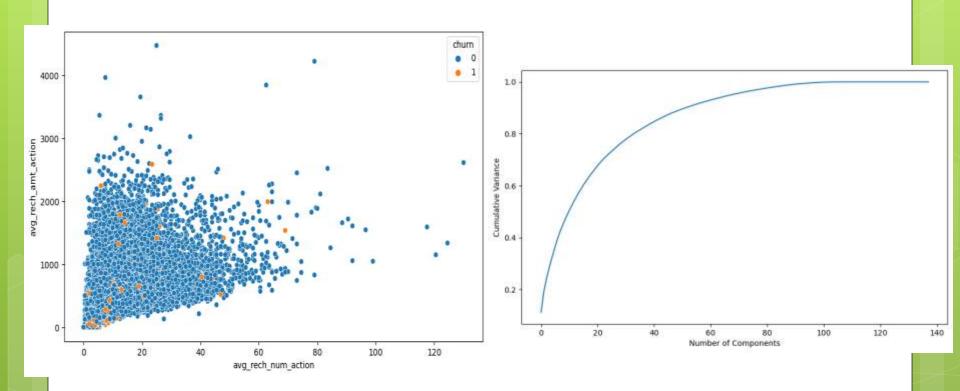


Observations:

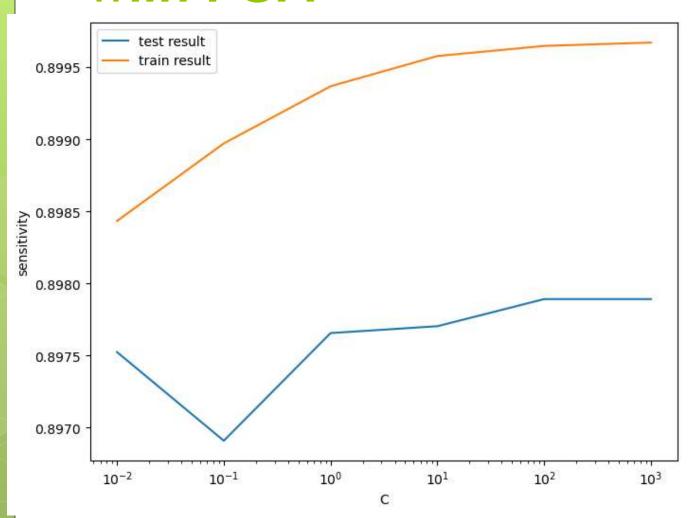
We can observe from the above plot that the churn rate is more for the customers, whose recharge amount as well as number of recharge have decreased in the action phase than the good phase.

Analysis of churn rate by the decreasing recharge amount and volume based cost in the action phase

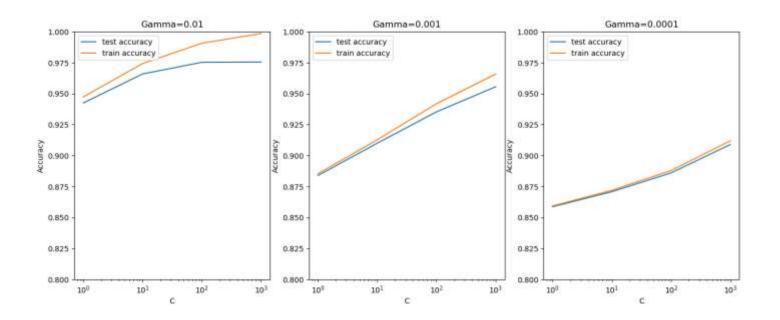
Step 7: Model Building with PCA



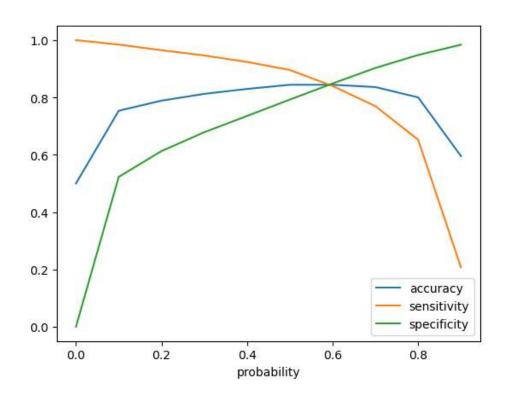
Step 8: Logistic regression with PCA



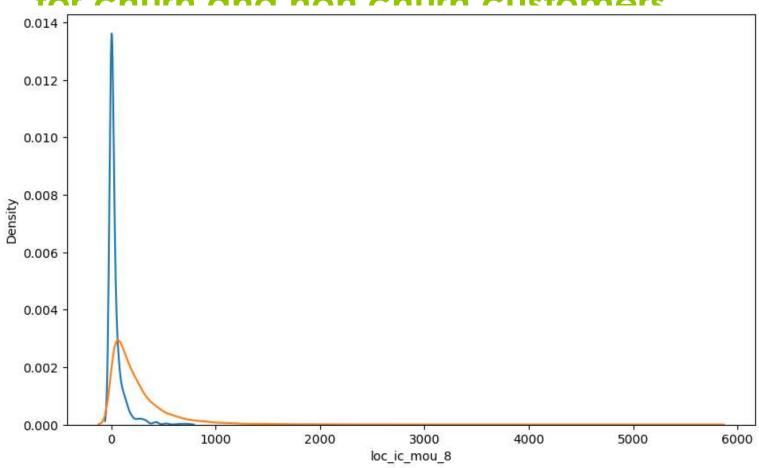
Step 9.2: Plotting the accuracy with various C and gamma values

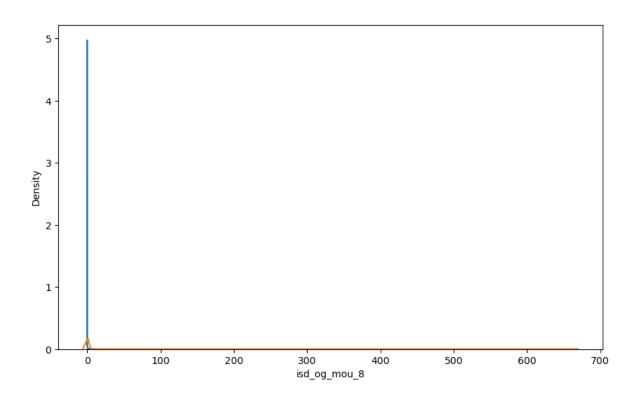


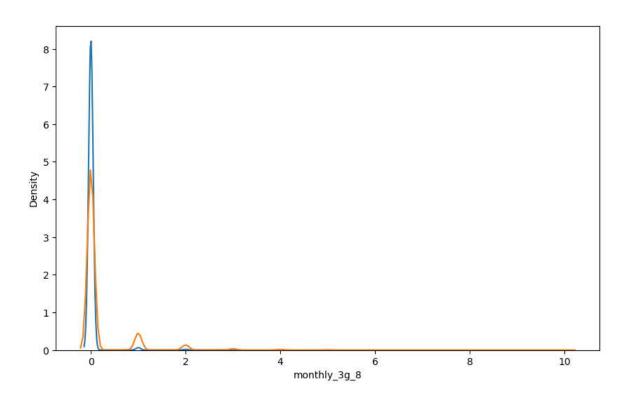
Step 15: Calculate the accuracy sensitivity and specificity for various probability cutoffs.



Step 18: Plots of important predictors for churp and non churp customers







THANK YOU