What is the Problem?

WSDM - KKBox's Churn Prediction Challenge

What is Churn?



Source:custify.com

Churn quantifies the number of customers who have left your brand by cancelling their subscription or stopping paying for your services.

Source:https://www.appier.com/blog/churn-prediction/

Business Impact

Since the music streaming service providers are becoming more competitive day by day one of the major problem these companies are facing is customer retention. High Churn Rate is bad news for any business as it costs five times as much to attract a new customer as it does to keep an existing one.

A high customer churn rate will hit your company's finances hard.



Prbolem Statement

To predict whether a user will churn after his/her subscription expires.

Specifically, we want to forecast if a user make a new service subscription transaction within 30 days after the current membership expiration date.

Client

KKBOX is Asia's leading music streaming service, holding the world's most comprehensive Asia-Pop music library with over 30 million tracks, supported by advertising and paid subscriptions.

USeful Links

- https://www.kaggle.com/c/kkbox-churn-prediction-challenge
- https://arxiv.org/ftp/arxiv/papers/1802/1802.03396.pdf · https://www.appier.com/blog/churn-prediction/
- **Business objective and constraints**

The company uses survival analysis techniques to determine the residual membership life time for each subscriber. By adopting different methods, KKBOX anticipates they'll discover new insights to why users leave Accurately predicting churn is critical to long-term success. Even slight variations in churn can drastically affect profits. No latency connstrains Minimum error Have probabilistic Output

Machine Learning Problem

Description

The churn/renewal definition can be tricky due to KKBox's subscription model. Since the majority of KKBox's subscription length is 30 days, a lot of users re-subscribe every month. The key fields to determine churn/renewal are transaction date, membership expiration date, and is_cancel. Note that the is_cancel field indicates whether a user actively cancels a subscription. Subscription cancellation does not imply the user has churned. A user may cancel service subscription due to change of service plans or other reasons. The criteria of "churn" is no new valid service subscription within 30 days after the current membership expires.

consists of users whose subscription expires within the month of February 2017, and the test data is with users whose subscription expires within the month of March 2017. This means we are looking at user churn or renewal roughly in the month of March 2017 for train set, and the user churn or renewal roughly in the month of April 2017. Train and test sets are split by transaction date,

The training and the test data are selected from users whose membership expire within a certain month. The train data



Data train_v2.csv

· msno: user id

• is churn: This is the target variable. Churn is defined as whether the user did not continue the subscription within 30 days of expiration. is_churn = 1 means churn,is_churn = 0 means renewal.

the train set, containing the user ids and whether they have churned.

of expiration. is_churn = 1 means churn,is_churn = 0 means renewal.

transactions.csv

sample_submission_v2.csv · msno: user id • is_churn: This is the target variable. Churn is defined as whether the user did not continue the subscription within 30 days

transactions of users up until 2/28/2017. · msno: user id

payment_method_id: payment method • payment_plan_days: length of membership plan in days plan_list_price: in New Taiwan Dollar (NTD)

- actual_amount_paid: in New Taiwan Dollar (NTD) is_auto_renew
- transaction_date: format %Y%m%d
- membership_expire_date: format %Y%m%d is_cancel: whether or not the user canceled the membership in this transaction.

msno: user id

user_logs.csv

· date: format %Y%m%d num_25: # of songs played less than 25% of the song length • num_50: # of songs played between 25% to 50% of the song length

daily user logs describing listening behaviors of a user. Data collected until 2/28/2017.

- num_75: # of songs played between 50% to 75% of of the song length
- num_985: # of songs played between 75% to 98.5% of the song length num_100: # of songs played over 98.5% of the song length num_unq: # of unique songs played total_secs: total seconds played
- members.csv user information. Note that not every user in the dataset is available.
- msno city

is_churn

registered_via: registration method registration_init_time: format %Y%m%d

waLDQMmcOu2jLDaV1ddDkgCrB/jl6sD66Xzs0Vqax1Y= QA7uiXy8vIbUSP0kCf9RwQ3FsT8jVq20xDr8zqa7bRQ=

• bd: age gender

Sample DataPoints

Train.csv

Members.csv msno registration_init_time 0 waLDQMmcOu2jLDaV1ddDkgCrB/jl6sD66Xzs0Vqax1Y=

fGwBva6hikQmTJzrbz/2Ezjm5Cth5jZUNvXigKK2AFA=

mT5V8rEpa+8wuqi6x0DoVd3H5icMKkE9Prt49UlmK+4=

XaPhtGLk/5UvvOYHcONTwsnH97P4eGECeq+BARGItRw=

20050406.0

1	QA7uiXy8vIbUSP0kCf9RwQ3FsT8jVq20xDr8zqa7bRQ=	10.0	38.0	male	9.0	
200	50407.0					
2	fGwBva6hikQmTJzrbz/2Ezjm5Cth5jZUNvXigKK2AFA=	11.0	27.0	female	9.0	2
005	1016.0					
3	mT5V8rEpa+8wuqi6x0DoVd3H5icMKkE9Prt49UlmK+4=	13.0	23.0	female	9.0	2
005	1102.0					
4	XaPhtGLk/5UvvOYHcONTwsnH97P4eGECeq+BARGItRw=	3.0 27.	0	male	9.0	
200	51228.0					

1

1

1

1

city

18.0

bd gender registered_via

36.0

female 9.0

Posing as a Machine Learning Problem

Binary class classification: is_churn either 0 or 1

Evaluration Metrics LogLoss: KPI Other metric to keep track

Confusion Matrix F1 Score

- **USeful Blogs And Reference**
- https://www.geeksforgeeks.org/python-working-with-date-and-time-using-pandas/ https://www.kaggle.com/c/kkbox-churn-prediction-challenge/discussion/46078 • https://www.kaggle.com/jeru666/did-you-think-of-these-features
- https://arxiv.org/ftp/arxiv/papers/1802/1802.03396.pdf Next

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Exploratory Data Analysis What do you want me to stay - EDA-KKBOX

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