# Credit card churning

We define that a credit card is churned by a customer if all cards of that customer become inactive for three consecutive months. You, as a data scientist, are assigned to build a model and predict customer credit card churning signals for the next three months (Jan 2018 - Mar 2018).

You are given the list of customers, their demographics, financial history and miscellaneous data. Below are the specification of your inputs and how we evaluate the results.

#### Few hints regarding user pool for this question.

- Every user has their credit card activated by the time the data was given
- Every user has transaction with a card at least once in year 2017
- Each user was limited to a maximum of three cards per person (Selecting ones with most transactions)

#### Provided Files:

• File name: y\_train.csv

o Output for selected user ids

o Size: 494 rows

| Field Name | Data Type | Description          |
|------------|-----------|----------------------|
| user_id    | INT       | Dummy user id        |
| Label      | INT       | 0—Active, 1—Inactive |

• File name: demo.csv

o Personal information of customers

o Size: 52,762 rows

| Field Name         | Data Type | Description              |
|--------------------|-----------|--------------------------|
| user_id            | INT       | Dummy user id            |
| account_start_date | TIMESTAMP | First account start date |

| birth_year                                 | INT    | Birth year                                                                         |
|--------------------------------------------|--------|------------------------------------------------------------------------------------|
| gender                                     | STRING | F-Female, M-Male                                                                   |
| marital_status                             | INT    | 1-Single, 2-Married, 3-<br>Divorce/widow                                           |
| <pre>family_income_segme nt_code</pre>     | INT    | Family income segment in ordinal scale (1 as lowest income segment)                |
| <pre>individual_income_s egment_code</pre> | INT    | <pre>Individual income segment in ordinal scale (1 as lowest income segment)</pre> |

• File name: card\_info.csv o Credit card information

o Size: 60,296 rows

| Field Name | Data Type | Description              |
|------------|-----------|--------------------------|
| user_id    | INT       | Dummy user id            |
| bill_cyc   | INT       | Bill cycle day of month  |
| cr_lmt_amt | INT       | Credit limit amount      |
| card_no    | INT       | Dummy credit card number |

• File name: cc\_txn.csv

o Credit card transaction log

o Period: 12 months o Size: 3,223,075 rows

| Field Name | Data Type | Description              |
|------------|-----------|--------------------------|
| mcc        | STRING    | Merchant category        |
| txn_dt     | TIMESTAMP | Transaction timestamp    |
| user_id    | INT       | Dummy user id            |
| txn_amt    | FLOAT     | Transaction amount       |
| card_no    | INT       | Dummy credit card number |

The category code descriptions are as follows:

| No. | Main Category                           | No. | Main Category |
|-----|-----------------------------------------|-----|---------------|
| 1   | Fashion and Apparel                     | 9   | Sports        |
| 2   | Health and beauty                       | 10  | Children      |
| 3   | Food and Beverage                       | 11  | Services      |
| 4   | Appliance and Electronics               | 12  | Education     |
| 5   | Office supplies,<br>books and gift shop | 13  | Pet           |
| 6   | Automotive shops and Vehicles           | 14  | Travel        |
| 7   | Entertainment                           | 15  | Accommodation |
| 8   | Home and Garden                         | 16  | Others        |

• File name: sa\_bal.csv

o Saving account balance aggregated by months

o Period: 12 months o Size: 633,144 rows

| Field Name | Data Type | Description                                 |
|------------|-----------|---------------------------------------------|
| user_id    | INT       | Dummy user id                               |
| mm         | INT       | Month of year                               |
| max_sa_bal | FLOAT     | Maximum saving account balance in the month |

• File name: dtxn.csv

o Incoming and outgoing transactions aggregated by months (exclude credit cardtransactions)

o Period: 12 months o Size: 490,599 rows

| Field Name | Data Type | Description              |
|------------|-----------|--------------------------|
| user_id    | INT       | Dummy user id            |
| mm         | INT       | Month of year            |
| amt_in     | FLOAT     | Amount of money inbound  |
| amt_out    | FLOAT     | Amount of money outbound |

### **Evaluation**

The problem focuses on anticipating the customers who are likely to stop using our credit cards. The business team will handle the rest to maintain those customers. Although precision is also important, the evaluation optimizes these two factors with higher emphasis on recall. Hence, we use F-2 score of label 1:

$$F_2 = 5 * \frac{precision * recall}{(4 * precision) + recall}$$

## **Objective**

Let you present data exploration and preparation Churn Prediction Challenge and build a predictive model to classify users between 2 classes: 0—Active and 1—Inactive based their activity and payment history along all their lifetime on service. And evaluate by using F-2 score of label 1 of the test set (use y\_train.csv split data 70/30).

Remark: Some of the data provided maybe incomplete, inconsistent, missing, noisy, erroneous, etc. as can occur in the real-world setting. It is the participants' task to recognize such cases as the challenge intentionally posed by the problem designer.