

# Renew and Return Rate Analysis

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```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

contracts = pd.read_excel('Data contracts.xlsx')
```

## Exploring Data

First, let's take a look at the head of data:

```
print(contracts.head())
```

	shop_id	package_order_id	package_name	contract_date	start_date	\
0	35120	57868	Car C	2021-08-22	2021-09-06	
1	73135	55723	Car B	2021-06-08	2021-06-08	
2	28746	49014	Car C	2021-01-16	2021-01-16	
3	76180	63743	Car C	2022-08-09	NaN	
4	63157	46291	Car B	2020-11-23	2020-11-23	

  

	end_date	real_end_date	Listing_limit	industry	category	region	\
0	2022-10-06	NaT	80	re_auto			
1	2022-07-15	NaT	50	re_auto			
2	2021-05-17	NaT	500	re_auto			
3	2023-02-08	NaT	10	re_auto			
4	2021-03-23	NaT	10	re_auto			

```
    city
0
1
2
3
4
```

It would be wise to change farsi columns into English, but because of time limit, I ignore this step.

## Data Types

```
print(contracts.info())
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14419 entries, 0 to 14418
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   shop_id               14419 non-null  int64
1   package_order_id     14419 non-null  int64
2   package_name         14419 non-null  object
3   contract_date        14417 non-null  object
4   start_date           14418 non-null  object
5   end_date             14419 non-null  object
6   real_end_date        8 non-null     datetime64[ns]
7   Listing_limit        14419 non-null  int64
8   industry             14419 non-null  object
9   category             14419 non-null  object
10  region               14416 non-null  object
11  city                 14419 non-null  object
dtypes: datetime64[ns](1), int64(3), object(8)
memory usage: 1.3+ MB
None
```

contract\_date, start\_date, and end\_date should be transformed into datetime objects.

**NOTE:** There are missing values in columns contract\_date, start\_date, and region.

```

contracts['contract_date'] = pd.to_datetime(contracts['contract_date'])

contracts['start_date'] = pd.to_datetime(contracts['start_date'])

contracts['end_date'] = pd.to_datetime(contracts['end_date'])

```

Checking date columns again:

```
print(contracts.info())
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 14419 entries, 0 to 14418
Data columns (total 12 columns):
#   Column                Non-Null Count  Dtype
---  -
0   shop_id               14419 non-null  int64
1   package_order_id     14419 non-null  int64
2   package_name         14419 non-null  object
3   contract_date        14417 non-null  datetime64[ns]
4   start_date           14418 non-null  datetime64[ns]
5   end_date             14419 non-null  datetime64[ns]
6   real_end_date        8 non-null      datetime64[ns]
7   Listing_limit        14419 non-null  int64
8   industry             14419 non-null  object
9   category             14419 non-null  object
10  region               14416 non-null  object
11  city                 14419 non-null  object
dtypes: datetime64[ns](4), int64(3), object(5)
memory usage: 1.3+ MB
None

```

We have also categorical column like industry, category, region, and city. In case of necessity, those columns will be defined as categories later.

### Unique values of each column

```
print(contracts.nunique())
```

```

shop_id            8123
package_order_id   14417
package_name        9
contract_date       1186
start_date          1130
end_date            1549
real_end_date        7
Listing_limit       48
industry            2
category            11
region              29
city                282
dtype: int64

```

There are 8123 unique values of shop\_id. Also, we have rows with the same package\_order\_id, which doesn't make a sense.

```

print(contracts[contracts.duplicated(subset='package_order_id', keep=False)].\
      sort_values('package_order_id'))

```

```

      shop_id  package_order_id  package_name  contract_date  start_date  \
9828      14546                5660    General C    2019-07-20  2019-07-20
11579      14546                5660    General C    2019-07-20  2019-07-20
280        74876                60314        Car B    2022-01-08  2022-01-08
11651      74876                60314        Car B    2022-02-07  2022-01-08

      end_date  real_end_date  Listing_limit  industry  category  \
9828  2019-10-21            NaT              5    re_auto
11579 2019-10-21            NaT              5    general
280    2022-07-09            NaT             10    re_auto
11651 2022-07-09            NaT             10    general

      region  city
9828
11579
280
11651

```

According to industry and category columns, these two orders are different from each other; however, their shop\_id, package\_name, start\_date and end\_date and even their city and region are identical!

There must be a mistake at data entry pipeline. Due to high uncertainty, all four rows are discarded from the following analysis.

```
contracts.drop_duplicates(subset='package_order_id', keep=False, inplace=True)
```

## Missing Values

```
missing_conditions = contracts['contract_date'].isna() | contracts['start_date'].isna() |
print(contracts[missing_conditions])
```

	shop_id	package_order_id	package_name	contract_date	start_date	\
3	76180	63743	Car C	2022-08-09	NaT	
10600	71562	53139	General A	2021-03-13	2021-03-13	
11147	68825	48980	General B	2021-01-13	2021-01-13	
12884	68392	48821	General C	2021-01-06	2021-01-06	
12955	76557	64920	General C	NaT	2022-10-26	
13769	76554	64903	General A	NaT	2022-10-26	

	end_date	real_end_date	Listing_limit	industry	\
3	2023-02-08	NaT	10	re_auto	
10600	2021-09-26	NaT	15	general	
11147	2021-04-14	NaT	5	general	
12884	2021-05-07	NaT	30	general	
12955	2023-01-24	NaT	5	general	
13769	2023-10-25	NaT	5	general	

	category	region	city
3			
10600		NaN	
11147		NaN	
12884		NaN	
12955			
13769			

NaT for date columns and NaN for other types are both standard ways of missingness indications.

## Month column

```
contracts['month'] = contracts['end_date'].dt.month
contracts['month'] = np.where(contracts['real_end_date'].notnull(), contracts['real_end_da
```

Months found in this data set:

```
print(contracts['month'].unique())
```

```
[10.  7.  5.  2.  3.  4.  8.  1. 11. 12.  6.  9.]
```

**Let's do the calculation for the last three months of year (i.e. October, November, and December).**

## Renew and Return Rate Calculation

First, a new column represents any future start\_date for a new contract:

```
contracts_sorted = contracts.sort_values(by = ['shop_id', 'start_date'])
contracts_sorted['start_date_next'] = contracts_sorted.groupby('shop_id')['start_date'].sh
```

Then, days between end\_date of previous contract and start\_date of new one is determined:

```
contracts_sorted['days_to_new'] = contracts_sorted['start_date_next'] - contracts_sorted['end_date']

# If real_end_date exists:
contracts_sorted['days_to_new'] = np.where(contracts_sorted['real_end_date'].notnull(), co

contracts_sorted['days_to_new'] = contracts_sorted['days_to_new'].dt.days
```

A new column represents whether the conditions of renewal or return have been met:

```
# Renew
contracts_sorted['renew'] = contracts_sorted['days_to_new'] <= 30
# Return
contracts_sorted['return'] = contracts_sorted['days_to_new'] > 30
```

Let's filter only those rows that its end\_date is within the last three months of year:

```
condition_months = contracts_sorted['month'].isin([10, 11, 12])
contracts_endseason = contracts_sorted[condition_months]
```

Last but not least, our interested rate are calculated:

```
renew_return_rates = contracts_endseason.groupby(['region', 'category', 'month'])['renew',
renew_return_rates = renew_return_rates.reset_index()
```

/var/folders/0s/5nvdy8kx70q7tvqqp\_3xvyx00000gn/T/ipykernel\_5150/4062478544.py:1: FutureWarning

Indexing with multiple keys (implicitly converted to a tuple of keys) will be deprecated, use

Renaming columns:

```
renew_return_rates.rename(columns = {'renew':'renew_rate_perc', 'return':'return_rate_perc
```

The dataframe is like the following:

```
print(renew_return_rates.head())
```

	region	category	month	renew_rate_perc	return_rate_perc
0		10.0		40.0	20.0
1		11.0		100.0	0.0
2		12.0		20.0	10.0
3		11.0	100.0		0.0
4		12.0	50.0		0.0