

```
In [1]: %pip install graphviz
```

Requirement already satisfied: graphviz in ./opt/anaconda3/lib/python3.9/site-packages (0.20.1)  
Note: you may need to restart the kernel to use updated packages.

```
In [2]: %pip install pydotplus
```

Requirement already satisfied: pydotplus in ./opt/anaconda3/lib/python3.9/site-packages (2.0.2)  
Requirement already satisfied: pyparsing>=2.0.1 in ./opt/anaconda3/lib/python3.9/site-packages (from pydotplus) (3.0.9)  
Note: you may need to restart the kernel to use updated packages.

## Спецификация

- 'churn' - факт оттока в текущем месяце (целевая переменная)
- 'gender' - пол
- 'near\_location' - проживание или работа в районе, где находится фитнес-центр
- 'partner' - сотрудник компании-партнёра клуба
- 'promo\_friends' - факт первоначальной записи в рамках акции "приведи друга"
- 'phone' - наличие контактного телефона
- 'age' - возраст
- 'lifetime' - время с первого обращения в фитнес-центр
- 'contract\_period' - длительность текущего действующего абонеента (месяц, 3 месяца, 6 месяцев, 1 год)
- 'month\_to\_end\_contract' - срок до окончания текущего действующего абонеента (в месяцах)
- 'group\_visits' - факт посещения групповых занятий
- 'avg\_class\_frequency\_total' - средняя частота посещений в неделю за все время с начала действия абонеента
- 'avg\_class\_frequency\_current\_month' - средняя частота посещений в неделю за предыдущий месяц
- 'avg\_additional\_charges\_total' - суммарная выручка от других услуг фитнес-центра: кафе, спорт-товары, косметический и массажный салон

## Предварительный анализ

### Загрузим необходимые библиотеки

```
In [47]: from io import StringIO
from IPython.display import Image
import graphviz
import pydotplus
import numpy as np
import pandas as pd
from typing import Dict, Tuple
from sklearn.datasets import load_iris, load_wine, load_boston
from sklearn.tree import DecisionTreeClassifier, DecisionTreeRegressor, export_graphviz
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.ensemble import RandomForestRegressor, GradientBoostingRegressor
```

```

from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
from sklearn.metrics import mean_absolute_error
from sklearn.metrics import median_absolute_error, r2_score
from sklearn.svm import NuSVR
from sklearn import tree
import matplotlib.pyplot as plt
%matplotlib inline

```

```
In [4]: df = pd.read_csv('gym.csv')
```

```
In [5]: df.head()
```

```
Out[5]:
```

	Unnamed: 0	gender	Near_Location	Partner	Promo_friends	Phone	Contract_period	Group_visits
0	0	1	1	1	1	0	6	1
1	1	0	1	0	0	1	12	1
2	2	0	1	1	0	1	1	1
3	3	0	1	1	1	1	12	1
4	4	1	1	1	1	1	1	1

## Изучим полученные данные

```
In [6]: df.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4000 entries, 0 to 3999
Data columns (total 15 columns):
 #   Column                                Non-Null Count  Dtype
---  -
 0   Unnamed: 0                          4000 non-null   int64
 1   gender                              4000 non-null   int64
 2   Near_Location                      4000 non-null   int64
 3   Partner                            4000 non-null   int64
 4   Promo_friends                      4000 non-null   int64
 5   Phone                              4000 non-null   int64
 6   Contract_period                    4000 non-null   int64
 7   Group_visits                       4000 non-null   int64
 8   Age                                4000 non-null   int64
 9   Avg_additional_charges_total       4000 non-null   float64
10   Month_to_end_contract              4000 non-null   float64
11   Lifetime                           4000 non-null   int64
12   Avg_class_frequency_total          4000 non-null   float64
13   Avg_class_frequency_current_month  4000 non-null   float64
14   Churn                              4000 non-null   int64
dtypes: float64(4), int64(11)
memory usage: 468.9 KB

```

```

In [7]: print('Отток:')
print(df[df['Churn']==1].describe().T)
print('\nОставшиеся клиенты:')
print(df[df['Churn']==0].describe().T)

```

ОТТОК:

	count	mean	std \
Unnamed: 0	1061.0	2018.887842	1158.746899
gender	1061.0	0.510839	0.500118
Near_Location	1061.0	0.768143	0.422217
Partner	1061.0	0.355325	0.478837
Promo_friends	1061.0	0.183789	0.387495
Phone	1061.0	0.902922	0.296204
Contract_period	1061.0	1.728558	2.132371
Group_visits	1061.0	0.268615	0.443448
Age	1061.0	26.989632	2.895163
Avg_additional_charges_total	1061.0	115.082899	77.696419
Month_to_end_contract	1061.0	1.662582	1.964593
Lifetime	1061.0	0.990575	1.110799
Avg_class_frequency_total	1061.0	1.474995	0.694705
Avg_class_frequency_current_month	1061.0	1.044546	0.770237
Churn	1061.0	1.000000	0.000000

	min	25%	50% \
Unnamed: 0	7.000000	1024.000000	2032.000000
gender	0.000000	0.000000	1.000000
Near_Location	0.000000	1.000000	1.000000
Partner	0.000000	0.000000	0.000000
Promo_friends	0.000000	0.000000	0.000000
Phone	0.000000	1.000000	1.000000
Contract_period	1.000000	1.000000	1.000000
Group_visits	0.000000	0.000000	0.000000
Age	18.000000	25.000000	27.000000
Avg_additional_charges_total	0.148205	50.629127	103.814686
Month_to_end_contract	1.000000	1.000000	1.000000
Lifetime	0.000000	0.000000	1.000000
Avg_class_frequency_total	0.000000	1.010771	1.491187
Avg_class_frequency_current_month	0.000000	0.421337	0.979445
Churn	1.000000	1.000000	1.000000

	75%	max
Unnamed: 0	3018.000000	3996.000000
gender	1.000000	1.000000
Near_Location	1.000000	1.000000
Partner	1.000000	1.000000
Promo_friends	0.000000	1.000000
Phone	1.000000	1.000000
Contract_period	1.000000	12.000000
Group_visits	1.000000	1.000000
Age	29.000000	38.000000
Avg_additional_charges_total	165.616858	425.535220
Month_to_end_contract	1.000000	12.000000
Lifetime	1.000000	9.000000
Avg_class_frequency_total	1.956438	3.478646
Avg_class_frequency_current_month	1.588576	3.540271
Churn	1.000000	1.000000

Оставшиеся клиенты:

	count	mean	std \
Unnamed: 0	2939.0	1992.500851	1153.550497
gender	2939.0	0.510037	0.499984
Near_Location	2939.0	0.873086	0.332933
Partner	2939.0	0.534195	0.498914
Promo_friends	2939.0	0.353522	0.478144
Phone	2939.0	0.903709	0.295040
Contract_period	2939.0	5.747193	4.716942
Group_visits	2939.0	0.464103	0.498795
Age	2939.0	29.976523	3.009933
Avg_additional_charges_total	2939.0	158.445715	99.801599

Month_to_end_contract	2939.0	5.283089	4.363522
Lifetime	2939.0	4.711807	3.874780
Avg_class_frequency_total	2939.0	2.024876	1.016006
Avg_class_frequency_current_month	2939.0	2.027882	1.018994
Churn	2939.0	0.000000	0.000000

	min	25%	50% \
Unnamed: 0	0.000000	994.500000	1986.000000
gender	0.000000	0.000000	1.000000
Near_Location	0.000000	1.000000	1.000000
Partner	0.000000	0.000000	1.000000
Promo_friends	0.000000	0.000000	0.000000
Phone	0.000000	1.000000	1.000000
Contract_period	1.000000	1.000000	6.000000
Group_visits	0.000000	0.000000	0.000000
Age	19.000000	28.000000	30.000000
Avg_additional_charges_total	0.171862	76.920993	149.881171
Month_to_end_contract	1.000000	1.000000	6.000000
Lifetime	0.000000	2.000000	4.000000
Avg_class_frequency_total	0.000000	1.283137	2.043252
Avg_class_frequency_current_month	0.000000	1.297021	2.046697
Churn	0.000000	0.000000	0.000000

	75%	max
Unnamed: 0	2991.500000	3999.000000
gender	1.000000	1.000000
Near_Location	1.000000	1.000000
Partner	1.000000	1.000000
Promo_friends	1.000000	1.000000
Phone	1.000000	1.000000
Contract_period	12.000000	12.000000
Group_visits	1.000000	1.000000
Age	32.000000	41.000000
Avg_additional_charges_total	224.448274	552.590740
Month_to_end_contract	10.000000	12.000000
Lifetime	6.000000	31.000000
Avg_class_frequency_total	2.732944	6.023668
Avg_class_frequency_current_month	2.740648	6.146783
Churn	0.000000	0.000000

## Вывод на основе отличий признаков у клиентов, попавших в отток и оставшихся:

- Распределение по полу не отличается
- Ближе живущие или работающие клиенты реже попадают в отток, что вполне логично
- Сотрудники компаний-партнеров реже перестают посещать фитнес-клуб чем остальные клиенты
- Клиенты, пришедшие по акции "Приведи друга" в два раза реже уходят в отток
- Наличие мобильного номера клиента не отличается в группах
- Средняя длительность договора полне логично больше у оставшихся пользователей, у большей части клиентов в оттоке срок договора составляет 1 месяц (скорее это говорит о том, что чаще клиенты заключают договор на 1 месяц).
- Оставшиеся пользователи чаще ходят на групповые занятия (хотя возможно, что в отток попали те, кто еще не успел начать ходить на них)
- У групп есть разница в среднем возрасте клиентов
- Есть небольшое различие в выручке от других услуг фитнес-центра

- Оставшееся время до окончания срока действия договора логично больше у оставшихся клиентов.
- У клиентов в оттоке меньше среднее количество посещений в неделю за всё время
- Количество посещений в предыдущий месяц в два раза ниже у клиентов в оттоке

## С использованием метода `train_test_split` разделим выборку на обучающую и тестовую.

Построим модель бинарной классификации пользователей, где целевой признак - факт оттока пользователя в следующем месяце.

```
In [8]: X = df.drop(['Churn'], axis = 1)
y = df['Churn']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, ran
scaler = StandardScaler()
X_train_st = scaler.fit_transform(X_train)
X_test_st = scaler.transform(X_test)
```

## Логическая регрессия

- Линейная регрессия предсказывает непрерывную величину, например, курс валюты, стоимость недвижимости и прочее.
- Логистическая регрессия используется для предсказания бинарной величины (да/нет), например, как в нашей задаче произойдет отток клиентов или нет

```
In [9]: lr_model = LogisticRegression(random_state=0)
lr_model.fit(X_train_st, y_train)
lr_predictions = lr_model.predict(X_test_st)
lr_probabilities = lr_model.predict_proba(X_test_st)[: ,1]
print('Метрики для модели логистической регрессии:')
print('accuracy_score: {}\nprecision_score: {}\nrecall_score:{}'.format(
    accuracy_score(y_test, lr_predictions),
    precision_score(y_test, lr_predictions),
    recall_score(y_test, lr_predictions)
))
```

```
Метрики для модели логистической регрессии:
accuracy_score: 0.92375
precision_score: 0.8586387434554974
recall_score:0.8282828282828283
```

```
In [10]: features = pd.DataFrame(lr_model.coef_.T, X.columns).reset_index()
features.columns = ['feature', 'coef']
features['coef'] = features['coef'].apply(lambda x: abs(x))
features = features.sort_values(by='coef', ascending=False)
print('\nКоэффициенты признаков в оптимальной функции логистической регрессии:')
print(features)
```

Коэффициенты признаков в оптимальной функции логистической регрессии:

	feature	coef
13	Avg_class_frequency_current_month	4.464397
11	Lifetime	3.847031
12	Avg_class_frequency_total	3.310909
8	Age	1.091190
6	Contract_period	0.719265
9	Avg_additional_charges_total	0.554970
10	Month_to_end_contract	0.530538
7	Group_visits	0.394999
4	Promo_friends	0.278891
3	Partner	0.079610
2	Near_Location	0.073833
0	Unnamed: 0	0.053905
1	gender	0.012632
5	Phone	0.006023

## SVM

```
In [11]: # SVM
scaler = StandardScaler().fit(X_train)
x_train_scaled = pd.DataFrame(scaler.transform(X_train), columns=X_train.columns)
x_test_scaled = pd.DataFrame(scaler.transform(X_test), columns=X_train.columns)
x_train_scaled.describe()
```

```
Out[11]:
```

	Unnamed: 0	gender	Near_Location	Partner	Promo_friends	Phone
<b>count</b>	3.200000e+03	3.200000e+03	3.200000e+03	3.200000e+03	3.200000e+03	3.200000e+03
<b>mean</b>	-3.629042e-17	-2.720046e-17	6.068063e-17	-4.127254e-16	3.320261e-16	-7.208471e-17
<b>std</b>	1.000156e+00	1.000156e+00	1.000156e+00	1.000156e+00	1.000156e+00	1.000156e+00
<b>min</b>	-1.747211e+00	-1.015114e+00	-2.395171e+00	-9.692234e-01	-6.756712e-01	-3.161734e-01
<b>25%</b>	-8.567368e-01	-1.015114e+00	4.175068e-01	-9.692234e-01	-6.756712e-01	3.162821e-01
<b>50%</b>	6.519114e-04	9.851108e-01	4.175068e-01	-9.692234e-01	-6.756712e-01	3.162821e-01
<b>75%</b>	8.615233e-01	9.851108e-01	4.175068e-01	1.031754e+00	1.480010e+00	3.162821e-01
<b>max</b>	1.733713e+00	9.851108e-01	4.175068e-01	1.031754e+00	1.480010e+00	3.162821e-01

```
In [12]: nusvr_05 = NuSVR(nu=0.7, gamma = 'scale')
nusvr_05.fit(X_train, y_train)
```

```
Out[12]: NuSVR(nu=0.7)
```

## Дерево решений

```
In [53]: dt_none = DecisionTreeRegressor(max_depth=3)
dt_none.fit(X_train, y_train)
```

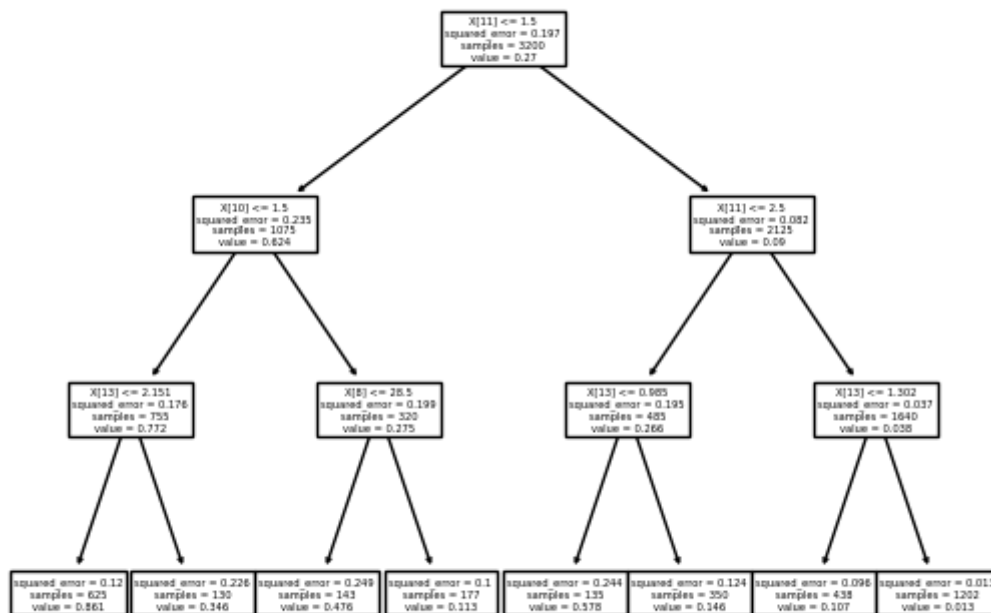
```
Out[53]: DecisionTreeRegressor(max_depth=3)
```

```
In [54]: tree.plot_tree(dt_none)
```

```

Out[54]: [Text(0.5, 0.875, 'X[11] <= 1.5\nsquared_error = 0.197\nsamples = 3200\nvalue = 0.27'),
          Text(0.25, 0.625, 'X[10] <= 1.5\nsquared_error = 0.235\nsamples = 1075\nvalue = 0.624'),
          Text(0.125, 0.375, 'X[13] <= 2.151\nsquared_error = 0.176\nsamples = 755\nvalue = 0.772'),
          Text(0.0625, 0.125, 'squared_error = 0.12\nsamples = 625\nvalue = 0.861'),
          Text(0.1875, 0.125, 'squared_error = 0.226\nsamples = 130\nvalue = 0.346'),
          Text(0.375, 0.375, 'X[8] <= 28.5\nsquared_error = 0.199\nsamples = 320\nvalue = 0.275'),
          Text(0.3125, 0.125, 'squared_error = 0.249\nsamples = 143\nvalue = 0.476'),
          Text(0.4375, 0.125, 'squared_error = 0.1\nsamples = 177\nvalue = 0.113'),
          Text(0.75, 0.625, 'X[11] <= 2.5\nsquared_error = 0.082\nsamples = 2125\nvalue = 0.09'),
          Text(0.625, 0.375, 'X[13] <= 0.985\nsquared_error = 0.195\nsamples = 485\nvalue = 0.266'),
          Text(0.5625, 0.125, 'squared_error = 0.244\nsamples = 135\nvalue = 0.578'),
          Text(0.6875, 0.125, 'squared_error = 0.124\nsamples = 350\nvalue = 0.146'),
          Text(0.875, 0.375, 'X[13] <= 1.302\nsquared_error = 0.037\nsamples = 1640\nvalue = 0.038'),
          Text(0.8125, 0.125, 'squared_error = 0.096\nsamples = 438\nvalue = 0.107'),
          Text(0.9375, 0.125, 'squared_error = 0.013\nsamples = 1202\nvalue = 0.013')]

```



```

In [22]: clf = DecisionTreeClassifier(random_state=1)
         clf.fit(X_train, y_train)

```

```

Out[22]: DecisionTreeClassifier(random_state=1)

```

```

In [19]: def test_model(model):
          print("mean_absolute_error:",
                mean_absolute_error(y_test, model.predict(X_test)))
          print("median_absolute_error:",
                median_absolute_error(y_test, model.predict(X_test)))
          print("r2_score:",
                r2_score(y_test, model.predict(X_test)))

```

```

In [17]: test_model(dt_none)

```

```
mean_absolute_error: 0.10875  
median_absolute_error: 0.0  
r2_score: 0.4160877881808114
```

```
In [21]: from IPython.core.display import HTML  
from sklearn.tree import export_text  
tree_rules = export_text(dt_none, feature_names=list(X.columns))  
HTML('<pre>' + tree_rules + '</pre>')
```



```

Out[21]: |--- Lifetime <= 1.50
|         |--- Contract_period <= 3.50
|         |         |--- Avg_class_frequency_current_month <= 2.15
|         |         |         |--- Age <= 32.50
|         |         |         |         |--- Avg_additional_charges_total <= 309.60
|         |         |         |         |         |--- Avg_class_frequency_current_month <= 1.26
|         |         |         |         |         |         |--- Avg_additional_charges_total <= 246.27
|         |         |         |         |         |         |         |--- Unnamed: 0 <= 10.50
|         |         |         |         |         |         |         |         |--- Avg_class_frequency_current_mon
th <= 1.05
|         |         |         |         |         |         |         |         |         |--- value: [1.00]
|         |         |         |         |         |         |         |         |         |--- Avg_class_frequency_current_mon
th > 1.05
|         |         |         |         |         |         |         |         |         |--- value: [0.00]
|         |         |         |         |         |         |         |         |         |--- Unnamed: 0 > 10.50
|         |         |         |         |         |         |         |         |         |--- Age <= 29.50
|         |         |         |         |         |         |         |         |         |--- Avg_additional_charges_tota
l <= 122.85
|         |         |         |         |         |         |         |         |         |--- value: [1.00]
|         |         |         |         |         |         |         |         |         |--- Avg_additional_charges_tota
l > 122.85
|         |         |         |         |         |         |         |         |         |--- Avg_additional_charges_
total <= 123.42
|         |         |         |         |         |         |         |         |         |--- value: [0.00]
|         |         |         |         |         |         |         |         |         |--- Avg_additional_charges_
total > 123.42
|         |         |         |         |         |         |         |         |         |--- truncated branch of
depth 5
|         |         |         |         |         |         |         |         |         |--- Age > 29.50
|         |         |         |         |         |         |         |         |         |--- Unnamed: 0 <= 2216.00
|         |         |         |         |         |         |         |         |         |--- Unnamed: 0 <= 2055.50
|         |         |         |         |         |         |         |         |         |--- truncated branch of
depth 5
|         |         |         |         |         |         |         |         |         |--- Unnamed: 0 > 2055.50
|         |         |         |         |         |         |         |         |         |--- value: [0.00]
|         |         |         |         |         |         |         |         |         |--- Unnamed: 0 > 2216.00
|         |         |         |         |         |         |         |         |         |--- value: [1.00]
|         |         |         |         |         |         |         |         |         |--- Avg_additional_charges_total > 246.27
|         |         |         |         |         |         |         |         |         |--- Avg_class_frequency_total <= 0.76
|         |         |         |         |         |         |         |         |         |--- Avg_class_frequency_current_mon
th <= 0.07
|         |         |         |         |         |         |         |         |         |--- value: [1.00]
|         |         |         |         |         |         |         |         |         |--- Avg_class_frequency_current_mon
th > 0.07
|         |         |         |         |         |         |         |         |         |--- Avg_additional_charges_tota
l <= 289.65
|         |         |         |         |         |         |         |         |         |--- value: [0.00]
|         |         |         |         |         |         |         |         |         |--- Avg_additional_charges_tota
l > 289.65
|         |         |         |         |         |         |         |         |         |--- value: [1.00]
|         |         |         |         |         |         |         |         |         |--- Avg_class_frequency_total > 0.76
|         |         |         |         |         |         |         |         |         |--- Avg_class_frequency_current_mon
th <= 1.08
|         |         |         |         |         |         |         |         |         |--- value: [1.00]
|         |         |         |         |         |         |         |         |         |--- Avg_class_frequency_current_mon
th > 1.08
|         |         |         |         |         |         |         |         |         |--- Avg_class_frequency_current

```

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49

4.18

```
| | | | | | | | | | | | |--- Unnamed: 0 > 699.00  
| | | | | | | | | | | | |--- value: [0.00]  
| | | | | | | | | | | | |--- Unnamed: 0 > 941.00  
| | | | | | | | | | | | |--- value: [1.00]  
| | | | | | | | | | | | |--- Avg_additional_charges_total > 14  
4.18  
| | | | | | | | | | | | |--- Avg_class_frequency_current_mon  
th <= 2.27  
| | | | | | | | | | | | |--- Partner <= 0.50  
| | | | | | | | | | | | |--- value: [1.00]  
| | | | | | | | | | | | |--- Partner > 0.50  
| | | | | | | | | | | | |--- value: [0.00]  
| | | | | | | | | | | | |--- Avg_class_frequency_current_mon  
th > 2.27  
| | | | | | | | | | | | |--- Avg_class_frequency_total <  
= 2.64  
| | | | | | | | | | | | |--- value: [0.00]  
| | | | | | | | | | | | |--- Avg_class_frequency_total >  
2.64  
| | | | | | | | | | | | |--- value: [1.00]  
| | | | | | | | | | | | |--- Avg_class_frequency_current_month > 2.  
58  
| | | | | | | | | | | | |--- Avg_class_frequency_total <= 2.48  
| | | | | | | | | | | | |--- value: [1.00]  
| | | | | | | | | | | | |--- Avg_class_frequency_total > 2.48  
| | | | | | | | | | | | |--- Avg_class_frequency_total <= 3.  
15  
| | | | | | | | | | | | |--- Avg_class_frequency_total <  
= 3.05  
| | | | | | | | | | | | |--- value: [0.00]  
| | | | | | | | | | | | |--- Avg_class_frequency_total >  
3.05  
| | | | | | | | | | | | |--- Group_visits <= 0.50  
| | | | | | | | | | | | |--- value: [0.00]  
| | | | | | | | | | | | |--- Group_visits > 0.50  
| | | | | | | | | | | | |--- value: [1.00]  
| | | | | | | | | | | | |--- Avg_class_frequency_total > 3.  
15  
| | | | | | | | | | | | |--- value: [1.00]  
| | | | | | | | | | | | |--- Unnamed: 0 > 1981.00  
| | | | | | | | | | | | |--- Avg_class_frequency_total <= 2.85  
| | | | | | | | | | | | |--- Avg_class_frequency_total <= 1.87  
| | | | | | | | | | | | |--- value: [1.00]  
| | | | | | | | | | | | |--- Avg_class_frequency_total > 1.87  
| | | | | | | | | | | | |--- Near_Location <= 0.50  
| | | | | | | | | | | | |--- Avg_class_frequency_total <  
= 2.38  
| | | | | | | | | | | | |--- value: [1.00]  
| | | | | | | | | | | | |--- Avg_class_frequency_total >  
2.38  
| | | | | | | | | | | | |--- value: [0.00]  
| | | | | | | | | | | | |--- Near_Location > 0.50  
| | | | | | | | | | | | |--- Avg_additional_charges_tota  
l <= 208.73  
| | | | | | | | | | | | |--- value: [0.00]  
| | | | | | | | | | | | |--- Avg_additional_charges_tota  
l > 208.73
```

localhost:8889/nbconvert/html/ML\_lab4.ipynb?download=false

```

| | | | |--- Month_to_end_contract > 7.50
| | | | |--- Avg_class_frequency_total <= 1.74
| | | | |--- Avg_additional_charges_total <= 82.53
11 | | | | |--- Avg_additional_charges_total <= 63.
| | | | |--- gender <= 0.50
| | | | |--- value: [1.00]
| | | | |--- gender > 0.50
| | | | |--- value: [0.00]
11 | | | | |--- Avg_additional_charges_total > 63.
| | | | |--- value: [1.00]
| | | | |--- Avg_additional_charges_total > 82.53
| | | | |--- value: [0.00]
| | | | |--- Avg_class_frequency_total > 1.74
| | | | |--- Group_visits <= 0.50
| | | | |--- value: [1.00]
| | | | |--- Group_visits > 0.50
| | | | |--- Promo_friends <= 0.50
| | | | |--- value: [1.00]
| | | | |--- Promo_friends > 0.50
| | | | |--- value: [0.00]
| | | | |--- Avg_class_frequency_current_month > 2.17
| | | | |--- Avg_additional_charges_total <= 10.85
| | | | |--- value: [1.00]
| | | | |--- Avg_additional_charges_total > 10.85
| | | | |--- Avg_class_frequency_total <= 2.50
| | | | |--- Avg_class_frequency_total <= 2.37
| | | | |--- Avg_class_frequency_current_month <
= 2.74
| | | | |--- Unnamed: 0 <= 3720.50
| | | | |--- value: [0.00]
| | | | |--- Unnamed: 0 > 3720.50
| | | | |--- value: [1.00]
2.74 | | | | |--- Avg_class_frequency_current_month >
| | | | |--- value: [1.00]
| | | | |--- Avg_class_frequency_total > 2.37
9.49 | | | | |--- Avg_additional_charges_total <= 16
| | | | |--- Avg_class_frequency_total <= 2.
42 | | | | |--- value: [1.00]
| | | | |--- Avg_class_frequency_total > 2.
42 | | | | |--- value: [0.00]
9.49 | | | | |--- Avg_additional_charges_total > 16
| | | | |--- value: [1.00]
| | | | |--- Avg_class_frequency_total > 2.50
| | | | |--- value: [0.00]
| | | |--- Age > 28.50
| | | |--- Avg_class_frequency_current_month <= 0.03
| | | |--- Avg_additional_charges_total <= 31.69
| | | |--- value: [0.00]
| | | |--- Avg_additional_charges_total > 31.69
| | | |--- value: [1.00]

```

```

| | | |--- Avg_class_frequency_current_month > 0.03
| | | |--- Lifetime <= 0.50
| | | |--- Avg_class_frequency_current_month <= 2.01
| | | |--- Avg_class_frequency_total <= 1.64
| | | |--- Avg_additional_charges_total <= 2.7
4
| | | |--- value: [1.00]
| | | |--- Avg_additional_charges_total > 2.7
4
| | | |--- Avg_class_frequency_current_mon
th <= 0.17
| | | |--- Age <= 29.50
| | | |--- value: [0.00]
| | | |--- Age > 29.50
| | | |--- value: [1.00]
| | | |--- Avg_class_frequency_current_mon
th > 0.17
| | | |--- value: [0.00]
| | | |--- Avg_class_frequency_total > 1.64
| | | |--- value: [1.00]
| | | |--- Avg_class_frequency_current_month > 2.01
| | | |--- Age <= 29.50
| | | |--- Avg_class_frequency_current_month <
= 2.72
| | | |--- value: [1.00]
| | | |--- Avg_class_frequency_current_month >
2.72
| | | |--- value: [0.00]
| | | |--- Age > 29.50
| | | |--- value: [0.00]
| | | |--- Lifetime > 0.50
| | | |--- Avg_additional_charges_total <= 57.65
| | | |--- Avg_class_frequency_current_month <= 1.
67
| | | |--- Avg_class_frequency_current_month <
= 1.06
| | | |--- value: [0.00]
| | | |--- Avg_class_frequency_current_month >
1.06
| | | |--- value: [1.00]
| | | |--- Avg_class_frequency_current_month > 1.
67
| | | |--- value: [0.00]
| | | |--- Avg_additional_charges_total > 57.65
| | | |--- Age <= 31.50
| | | |--- Avg_class_frequency_current_month <
= 2.08
| | | |--- Avg_class_frequency_total <= 2.
10
| | | |--- Avg_additional_charges_tota
l <= 86.57
| | | |--- Unnamed: 0 <= 1294.00
| | | |--- value: [1.00]
| | | |--- Unnamed: 0 > 1294.00
| | | |--- value: [0.00]
| | | |--- Avg_additional_charges_tota
l > 86.57

```



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```
| | | | | | | | | |--- Promo_friends > 0.50  
| | | | | | | | | |--- value: [1.00]  
| | | | | | | | | |--- Avg_additional_charges_total > 12  
8.67  
| | | | | | | | | |--- value: [0.00]  
| | | | | | | | | |--- Month_to_end_contract > 4.50  
| | | | | | | | | |--- value: [0.00]  
| | | | | | | | | |--- Avg_class_frequency_total > 0.94  
| | | | | | | | | |--- Avg_class_frequency_current_month <= 0.96  
| | | | | | | | | |--- value: [1.00]  
| | | | | | | | | |--- Avg_class_frequency_current_month > 0.96  
| | | | | | | | | |--- Group_visits <= 0.50  
| | | | | | | | | |--- value: [1.00]  
| | | | | | | | | |--- Group_visits > 0.50  
| | | | | | | | | |--- value: [0.00]  
| | | | | | | | | |--- Avg_class_frequency_current_month > 0.98  
| | | | | | | | | |--- Month_to_end_contract <= 2.00  
| | | | | | | | | |--- Age <= 27.50  
| | | | | | | | | |--- Avg_class_frequency_current_month <= 2.17  
| | | | | | | | | |--- Avg_class_frequency_total <= 1.72  
| | | | | | | | | |--- Group_visits <= 0.50  
| | | | | | | | | |--- Unnamed: 0 <= 2212.00  
| | | | | | | | | |--- value: [1.00]  
| | | | | | | | | |--- Unnamed: 0 > 2212.00  
| | | | | | | | | |--- value: [0.00]  
| | | | | | | | | |--- Group_visits > 0.50  
| | | | | | | | | |--- Avg_class_frequency_total <= 1.  
02  
| | | | | | | | | |--- value: [1.00]  
| | | | | | | | | |--- Avg_class_frequency_total > 1.  
02  
| | | | | | | | | |--- value: [0.00]  
| | | | | | | | | |--- Avg_class_frequency_total > 1.72  
| | | | | | | | | |--- value: [1.00]  
| | | | | | | | | |--- Avg_class_frequency_current_month > 2.17  
| | | | | | | | | |--- Avg_additional_charges_total <= 56.52  
| | | | | | | | | |--- value: [1.00]  
| | | | | | | | | |--- Avg_additional_charges_total > 56.52  
| | | | | | | | | |--- gender <= 0.50  
| | | | | | | | | |--- Avg_additional_charges_total <=  
115.69  
| | | | | | | | | |--- value: [0.00]  
| | | | | | | | | |--- Avg_additional_charges_total >  
115.69  
| | | | | | | | | |--- value: [1.00]  
| | | | | | | | | |--- gender > 0.50  
| | | | | | | | | |--- value: [0.00]  
| | | | | | | | | |--- Age > 27.50  
| | | | | | | | | |--- Avg_class_frequency_current_month <= 1.85  
| | | | | | | | | |--- Avg_class_frequency_total <= 1.82  
| | | | | | | | | |--- Avg_additional_charges_total <= 10  
7.02  
| | | | | | | | | |--- Avg_additional_charges_total <=  
86.30  
| | | | | | | | | |--- Near_Location <= 0.50  
| | | | | | | | | |--- Age <= 31.00  
| | | | | | | | | |--- value: [1.00]
```

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51  
|  
|  
51

```
| | | | | | --- Unnamed: 0 > 398.00  
| | | | | | |--- Near_Location <= 0.50  
| | | | | | |--- Avg_class_frequency_current_month <  
= 2.05  
| | | | | | |--- value: [1.00]  
| | | | | | |--- Avg_class_frequency_current_month >  
2.05  
| | | | | | |--- value: [0.00]  
| | | | | | |--- Near_Location > 0.50  
| | | | | | |--- value: [0.00]  
| | | | | |--- Age > 24.50  
| | | | | |--- Lifetime <= 3.50  
| | | | | |--- Partner <= 0.50  
| | | | | |--- Unnamed: 0 <= 3844.00  
| | | | | | |--- Avg_class_frequency_current_mon  
th <= 1.93  
| | | | | | |--- Avg_class_frequency_total <  
= 2.08  
| | | | | | |--- Avg_class_frequency_cur  
rent_month <= 1.83  
| | | | | | |--- value: [0.00]  
| | | | | | |--- Avg_class_frequency_cur  
rent_month > 1.83  
| | | | | | |--- truncated branch of  
depth 2  
| | | | | | |--- Avg_class_frequency_total >  
2.08  
| | | | | | |--- value: [1.00]  
| | | | | | |--- Avg_class_frequency_current_mon  
th > 1.93  
| | | | | | |--- Avg_additional_charges_tota  
l <= 23.62  
| | | | | | |--- Avg_additional_charges_  
total <= 19.58  
| | | | | | |--- value: [0.00]  
| | | | | | |--- Avg_additional_charges_  
total > 19.58  
| | | | | | |--- value: [1.00]  
| | | | | | |--- Avg_additional_charges_tota  
l > 23.62  
| | | | | | |--- Unnamed: 0 <= 2737.50  
| | | | | | |--- value: [0.00]  
| | | | | | |--- Unnamed: 0 > 2737.50  
| | | | | | |--- truncated branch of  
depth 2  
| | | | | | |--- Unnamed: 0 > 3844.00  
| | | | | | |--- Avg_class_frequency_total <= 2.  
65  
| | | | | | |--- Month_to_end_contract <= 3.  
50  
| | | | | | |--- value: [0.00]  
| | | | | | |--- Month_to_end_contract > 3.  
50  
| | | | | | |--- Unnamed: 0 <= 3953.50  
| | | | | | |--- value: [1.00]  
| | | | | | |--- Unnamed: 0 > 3953.50  
| | | | | | |--- value: [0.00]
```

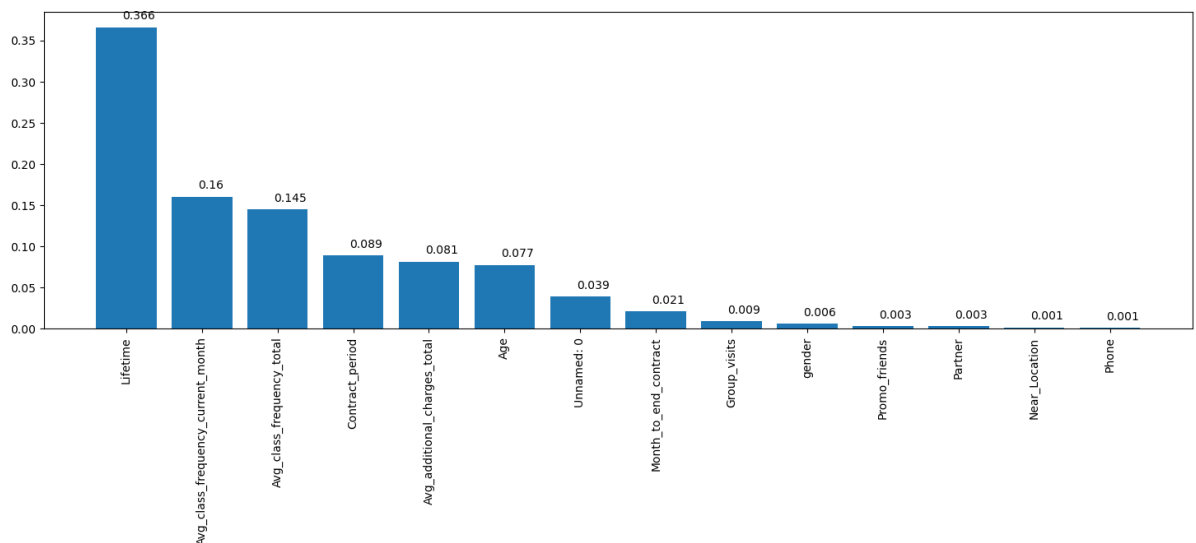


```
Out[41]: [('Unnamed: 0', 0.03890444087977608),
('gender', 0.00583693983884679),
('Near_Location', 0.0012667344323994687),
('Partner', 0.002845916239263289),
('Promo_friends', 0.002961742125806938),
('Phone', 0.0007880349584736334),
('Contract_period', 0.08862262628999662),
('Group_visits', 0.008962379294846418),
('Age', 0.07710914767980905),
('Avg_additional_charges_total', 0.08100038165972405),
('Month_to_end_contract', 0.020922651769533434),
('Lifetime', 0.3659062244517595),
('Avg_class_frequency_total', 0.14481449149670683),
('Avg_class_frequency_current_month', 0.1600582888830579)]
```

```
In [42]: from operator import itemgetter

def draw_feature_importances(tree_model, X_dataset, figsize=(18,5)):
    """
    Вывод важности признаков в виде графика
    """
    # Сортировка значений важности признаков по убыванию
    list_to_sort = list(zip(X_dataset.columns.values, tree_model.feature_imp
sorted_list = sorted(list_to_sort, key=itemgetter(1), reverse = True)
    # Названия признаков
    labels = [x for x, _ in sorted_list]
    # Важности признаков
    data = [x for _, x in sorted_list]
    # Вывод графика
    fig, ax = plt.subplots(figsize=figsize)
    ind = np.arange(len(labels))
    plt.bar(ind, data)
    plt.xticks(ind, labels, rotation='vertical')
    # Вывод значений
    for a,b in zip(ind, data):
        plt.text(a-0.05, b+0.01, str(round(b,3)))
    plt.show()
    return labels, data
```

```
In [43]: dt_fl, dt_fd = draw_feature_importances(clf, X_train)
```



```
In [ ]:
```