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""2022 г.	""	2022 г.			
Отчет по рубежному контрол Технологии машинного ГУИМЦ		rypcy			
Тема работы: '' Методы построен обучения.		й машинного			
16 (количество листов) <u>Вариант № 28</u>)				
ИСПОЛНИТЕЛЬ:					
студент группы ИУ5Ц-82Б					
Чиварзин А.Е.	""	дпись) 2022 г.			

Цель лабораторной работы

Изучение линейных моделей, SVM и деревьев решений

Задание

- 1. Выберите набор данных (датасет) для решения задачи классификации или регрессии.
- 2. В случае необходимости проведите удаление или заполнение пропусков и кодирование категориальных признаков.
- 3. С использованием метода train test split разделите выборку на обучающую и тестовую.
- 4. Обучите следующие модели:
 - одну из линейных моделей (линейную или полиномиальную регрессию при решении задачи регрессии, логистическую регрессию при решении задачи классификации);
 - SVM:
 - дерево решений.
- 5. Оцените качество моделей с помощью двух подходящих для задачи метрик. Сравните качество полученных моделей.
- 6. Постройте график, показывающий важность признаков в дереве решений.
- 7. Визуализируйте дерево решений или выведите правила дерева решений в текстовом виде.

Ход лабораторной работы

Текстовое описание набора данных

В качестве набора данных используется dataset рейтингов университетов мира на основании трёх рейтингов. Датасет доступен по адресу: https://www.kagqle.com/mylesoneill/world-university-rankings

Из набора данных будет рассматриваться только файл cwurData.csv.

Описание столбцов:

- world rank мировой рейтинг университета
- institution название университета
- country страна, в которой расположен университет
- national_rank рейтинг университета в стране его нахождения
- quality of education рейтинг качества образования
- quality of faculty рейтинг качества профессорско-преподавательского состава
- publications рейтинг публикаций
- infuence рейтинг влияния
- citations количество студентов в университете
- broad impact рейтинг за широкое влияние (предоставлен только за 2014 и 2015 гг. Остальное пропуски)
- patents рейтинг за патенты
- score общий балл, используемый для определения мирового рейтинга
- year год рейтинга (c 2012 по 2015 год)

Основные характеристики набора данных

Подключаем все необходимые библиотеки

```
In [1]: import numpy as np
    import pandas as pd
    import seaborn as sns
    import matplotlib
    import matplotlib_inline
    import matplotlib.pyplot as plt
    from sklearn.model_selection import train_test_split
    %matplotlib inline
    sns.set(style="ticks")
    from io import StringIO

Подключаем Dataset
In [2]: data = pd.read_csv('cwurData.csv', sep=",")
Размер набора данных
In [3]: data.shape
Out[3]: (2200, 14)
```

```
Типы колонок
```

In [4]: data.dtypes

```
Out[4]:world_rank int64
institution object
country object
national_rank int64
quality_of_education
alumni_employment int64
quality_of_faculty int64
publications int64
influence int64
citations int64
broad_impact float64
patents int64
score float64
year int64
dtype: object
```

Проверяем, есть ли пропущенные значения

In [5]: data.isnull().sum()

```
Out[5]:world_rank
                                0
     institution
                              0
     country
    quality_of_education 0
alumni_employment 0
quality_of_faculty 0
publications
     national rank
     influence
                              0
     citations
     broad_impact
                             200
                               0
     patents
     score
                               0
     year
                               0
```

dtype: int64 Первые 5 строк датасета

In [6]: data.head()

Out[6]:	world_rank	institution	country	national_rank	quality_of_education	alumni_employment	quality_of_faculty	publications	influence
0	1	Harvard University	USA	1	7	9	1	1	1
1	2	Massachusetts Institute of Technology	USA	2	9	17	3	12	4
2	3	Stanford University	USA	3	17	11	5	4	2
3	4	University of Cambridge	United Kingdom	1	10	24	4	16	16
4	5	California Institute of Technology	USA	4	2	29	7	37	22
4									Þ

```
In [7]: total_count = data.shape[0]
    print('Bcero ctpok: {}'.format(total_count))
Bcero ctpok: 2200
Προμεн προπусков в broad_impact
```

In [8]: (200 / 2200) * 100
Out[8]:9.090909090909092

Настройка отображения графиков

```
In [9]: # Задание формата графиков для сохранения высокого качества PNG from IPython.display import set_matplotlib_formats matplotlib_inline.backend_inline.set_matplotlib_formats("retina") # Задание ширины графиков, чтобы они помещались на A4 pd.set_option("display.width", 70)
```

Обработка пропусков данных

Очистка строк

Можно очистить строки, содержащие пропуски. При этом останутся данные только за 2014 и 2015 гг (см. описание датасета)

Out[10]: ((2200, 14), (2000, 14))

Выведем первые 11 строк, чтобы убедиться, что данные в national_rank числовые (Jupyter Lab в предпросмотре CSV показывает не совсем верно)

In [11]: data_no_null.head(11)

Out[11]:	world_rank	institution	country	national_rank	quality_of_education	alumni_employment	quality_of_faculty	publications	influenc
200	1	Harvard University	USA	1	1	1	1	1	
201	2	Stanford University	USA	2	11	2	4	5	
202	3	Massachusetts Institute of Technology	USA	3	3	11	2	15	
203	4	University of Cambridge	United Kingdom	1	2	10	5	10	
204	5	University of Oxford	United Kingdom	2	7	12	10	11	
205	6	Columbia University	USA	4	13	8	9	14	•
206	7	University of California, Berkeley	USA	5	4	22	6	7	
207	8	University of Chicago	USA	6	10	14	8	17	•
208	9	Princeton University	USA	7	5	16	3	70	í
209	10	Yale University	USA	8	9	25	11	18	
210	11	Cornell University	USA	9	12	18	19	23	

In [12]: total_count = data_no_null.shape[0]
 print('Bcero строк: {}'.format(total_count))

Всего строк: 2000

Кодирование категориальных признаков

Преобразуем названия стран, городов, ... в числовые зеачения (label encoding)

```
In [13]: from sklearn.preprocessing import LabelEncoder, OneHotEncoder
```

```
/tmp/ipykernel_143/4210865855.py:4: SettingWithCopyWarning: A value is trying to be set on a copy of a slice from a DataFrame. Try using .loc[row_indexer,col_indexer] = value instead
```

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#re turning-a-view-versus-a-copy

data_no_null.institution = le.transform(data_no_null.institution)

/tmp/ipykernel 143/4210865855.py:7: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame.

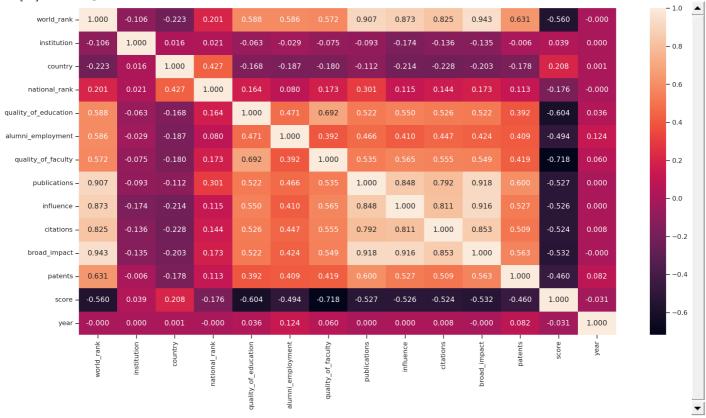
Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: $https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy$

data no null["country"] = le.transform(data no null["country"])

Построим кореляционную матрицу

Out[15]:<AxesSubplot:>



Предсказание целевого признака

Предскажем значение целевого признака $world_rank$ по $broad_impact$ и publications, поскольку их значения кореляции ближе всего к 1

Разбиение выборки на обучающую и тестовую

```
In [16]: X = data_no_null[["broad_impact", "publications"]]
    Y = data_no_null["world_rank"]
    print('Входные данные:\n\n', X.head(), '\n\nВыходные данные:\n\n', Y.head())
```

Входные данные:

	broad_impact	publications
200	1.0	1
201	4.0	5
202	2.0	15
203	13.0	10
204	12.0	11

Выходные данные:

Name: world_rank, dtype: int64

Разделим выборку на обучающую и тестовую

In [17]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, random_state = 2022, test_size = 0.1)

Входные параметры обучающей выборки

In [18]: X_train.head()

Out[18]:	broad_impact	publications
2164	932.0	875
1710	590.0	576
428	164.0	200
1389	164.0	233
2089	932.0	675

Входные параметры тестовой выборки

In [19]: $X_{test.head}()$

Out[19]:		$broad_impact$	publications
12	18	14.0	3
14	95	265.0	236
8	43	703.0	943
20	42	850.0	803
18	69	606.0	701

Выходные параметры обучающей выборки

In [20]: Y_train.head()

Out[20]:2164 965 1710 511 428 229 1389 190 2089 890

Name: world_rank, dtype: int64

Выходные параметры тестовой выборки

In [21]: Y_test.head()

Out[21]:1218 19 1495 296 843 644 2042 843 1869 670

Name: world_rank, dtype: int64

Построение линейной регрессии

In [22]: from sklearn.linear_model import LinearRegression
 from sklearn.metrics import mean_absolute_error, mean_squared_error, median_absolute_error, r2_score
In [23]: Lin_Reg = LinearRegression().fit(X_train, Y_train)
 lr_y_pred = Lin_Reg.predict(X_test)

Возьмем тот параметр, чья корреляция ближе всего к единице, т.е. broad_impact

```
In [40]: plt.scatter(X test["broad impact"], Y test,
                                                       marker = 's', label = 'Тестовая выборка')
      plt.scatter(X test["broad impact"], lr y pred, marker = 'o', label = 'Предсказанные данные')
      plt.legend (loc = 'lower right')
      plt.xlabel ('Рейтинг за широкое влияние')
      plt.ylabel ('Целевой признак')
      plt.show()
   1000
    800
Целевой признак
    600
    400
    200
                                Тестовая выборка
                                Предсказанные данные
                  Рейтинг за широкое влияние
In [25]: from sklearn.metrics import mean_absolute_error, mean_squared_error, median_absolute_error, r2_score
In [26]: print ('Средняя абсолютная ошибка:',
                                              mean absolute error (Y test, lr y pred))
      print('Средняя квадратичная ошибка:', mean_squared_error(Y_test, lr_y_pred))
                                              median absolute error(Y test, lr_y_pred))
      print('Median absolute error:',
      print ('Коэффициент детерминации:',
                                              r2 score(Y test, lr y pred))
Средняя абсолютная ошибка: 54.70203008487861
Средняя квадратичная ошибка: 6228.270901286782
Median absolute error: 42.05433711920929
Коэффициент детерминации: 0.9234584275958889
SVM
In [27]: {	t from} sklearn.svm {	t import} SVC , LinearSVC
      from sklearn.datasets import make blobs
In [28]: svc = SVC(kernel='linear')
      svc.fit(X_train,Y_train)
Out[28]:SVC(kernel='linear')
In [29]: pred_y = svc.predict(X_test)
ln [41]: plt.scatter(X_test["broad_impact"], Y_test, marker = 's', label = 'Тестовая выборка')
      plt.scatter(X test["broad impact"], pred y, marker = 'o', label = 'Предсказанные данные')
      plt.legend (loc = 'lower right')
      plt.xlabel ('рейтинг за широкое влияние')
      plt.ylabel ('Целевой признак')
      plt.show()
   1000
    800
Целевой признак
    600
    400
    200
                                Тестовая выборка
                                Предсказанные данные
                                 600
                200
                        400
                                         800
                                                 1000
                  рейтинг за широкое влияние
In [31]: print ('Средняя абсолютная ошибка:',
                                              mean_absolute_error(Y_test, pred_y))
      print('Средняя квадратичная ошибка:', mean squared error(Y test, pred y))
      print('Median absolute error:',
                                              median absolute error(Y test, pred y))
      print ('Коэффициент детерминации:',
                                              r2 score (Y test, pred y))
Средняя абсолютная ошибка: 57.19
Средняя квадратичная ошибка: 9379.4
Median absolute error: 27.0
Коэффициент детерминации: 0.8847330124868531
```

Дерево (Tree)

```
In [32]: from sklearn.tree import DecisionTreeClassifier, DecisionTreeRegressor, export_graphviz
      from sklearn.tree import export graphviz
      from sklearn import tree
      import re
      from IPython.core.display import HTML
      from sklearn.tree import export text
      import graphviz
      from IPython.display import Image
      import pydotplus
Обучим дерево на всех признаках
In [33]: reg = tree.DecisionTreeRegressor()
      reg = reg.fit(X test, Y test)
In [42]: pred_y = reg.predict(X_test)
      plt.scatter(X_test["broad_impact"], Y_test, marker = 's', label = 'Тестовая выборка')
      plt.scatter(X_test["broad_impact"], pred_y, marker = 'o', label = 'Предсказанные данные')
      plt.legend (loc = 'lower right')
      plt.xlabel ('рейтинг за широкое влияние')
      plt.ylabel ('Целевой признак')
      plt.show()
  1000
   800
Целевой признак
   600
    400
    200
                                Тестовая выборка
                                Предсказанные данные
                                 600
                                         800
                                                 1000
                200
                        400
                  рейтинг за широкое влияние
```

Дерево в текстовом виде

```
In [35]: tree_rules = export_text(reg, feature_names=list(X.columns))
     HTML('' + tree rules + '')
Out[35]: |--- broad impact <= 477.00
         |--- broad impact <= 201.50
             |--- publications <= 105.00
                  |--- publications <= 24.50
                      |--- publications <= 23.50
                          |--- publications <= 5.00
                          | |--- value: [19.00]
                           --- publications > 5.00
                             |--- broad impact <= 18.00
                                 |--- broad impact <= 8.00
                                 | |--- value: [7.00]
                                 |--- broad impact > 8.00
                                 | |--- publications <= 10.50
                                 | | |--- value: [5.00]
                                   |--- publications > 10.50
                                 | |--- value: [6.00]
                                 |--- broad_impact > 18.00
                             | |--- value: [11.00]
                      |--- publications > 23.50
                     | |--- value: [27.00]
                     - publications > 24.50
                      |--- broad impact <= 69.50
                          |--- broad impact <= 52.00
                             |--- publications <= 30.00
                             | |--- value: [53.00]
                                - publications > 30.00
                                 |--- broad impact <= 43.50
                                   |--- value: [69.00]
                                 |--- broad impact > 43.50
                                     |--- publications <= 65.50
```

```
| |--- value: [82.00]
          | | |--- publications > 65.50
| | | |--- value: [70.00]
        |--- broad_impact > 52.00
        | |--- publications <= 46.50
          | |--- value: [28.00]
       | |--- publications > 46.50
          | |--- value: [51.00]
       - broad impact > 69.50
        |--- publications <= 73.00
          |--- broad impact <= 75.00
           | |--- value: [93.00]
           |--- broad impact > 75.00
           | |--- publications <= 66.00
               | |--- value: [116.00]
           | |--- publications > 66.00
          | | |--- value: [127.00]
        |--- publications > 73.00
           |--- broad impact <= 139.00
               |--- broad impact <= 103.50
               | |--- value: [70.00]
            | |--- broad impact > 103.50
            | | |--- value: [43.00]
            |--- broad impact > 139.00
               |--- publications <= 82.50
               | |--- value: [87.00]
              |--- publications > 82.50
          | | |--- value: [145.00]
- publications > 105.00
|--- broad impact <= 63.00
  |--- value: [37.00]
|--- broad impact > 63.00
    |--- publications <= 155.50
        |--- publications <= 138.50
           |--- publications <= 120.50
               |--- publications <= 115.00
               | |--- value: [147.00]
            | |--- publications > 115.00
            | | |--- value: [157.00]
           |--- publications > 120.50
               |--- publications <= 133.00
                    |--- broad impact <= 130.50
                  | |--- value: [175.00]
                    |--- broad_impact > 130.50
                  | |--- publications <= 127.00
                   | | |--- value: [178.00]
                  | |--- publications > 127.00
| | |--- value: [179.00]
           | |--- publications > 133.00
          | | |--- value: [208.00]
        |--- publications > 138.50
          |--- publications <= 145.00
               |--- value: [146.00]
           |--- publications > 145.00
            | |--- value: [133.00]
       - publications > 155.50
        |--- broad impact <= 165.50
            |--- publications <= 190.50
            | |--- publications <= 173.00
            | | |--- value: [165.00]
            | |--- publications > 173.00
            | | |--- value: [168.00]
            |--- publications > 190.50
               |--- publications <= 220.50
               | |--- value: [197.00]
               |--- publications > 220.50
               | |--- publications <= 262.50
                  | |--- value: [194.00]
                    |--- publications > 262.50
                  | |--- value: [193.00]
        |--- broad impact > 165.50
            |--- publications <= 202.50
               |--- broad impact <= 175.00
               | |--- value: [205.00]
               |--- broad impact > 175.00
               | |--- broad impact <= 188.00
                    | |--- value: [216.00]
```

```
|--- publications > 202.50
                   |--- broad impact <= 185.00
                  | |--- value: [240.00]
                | |--- broad impact > 185.00
       | | | | |--- value: [229.00]
-- broad impact > 201.50
|--- broad impact <= 380.50
    |--- publications <= 399.00
       |--- publications <= 382.00
         | |--- publications <= 277.00
               |--- broad_impact <= 368.00
                   |--- broad impact <= 355.00
                      |--- publications <= 197.50
                   | |--- publications <= 100.00
                       | | |--- value: [191.00]
                       | |--- publications > 100.00
                       | | |--- broad_impact <= 276.50
                              | |--- truncated branch of depth 3
                             |--- broad_impact > 276.50
                          | | |--- value: [217.00]
                      |--- publications > 197.50
                         |--- broad impact <= 253.50
                             |--- broad impact <= 212.00
                              | |--- value: [237.00]
                             |--- broad impact > 212.00
                         | | |--- truncated branch of depth 3
                   | | |--- broad_impact > 253.50
                      | | |--- broad_impact <= 260.50
                   | |--- value: [313.00]
                             |--- broad impact > 260.50
                          | | |--- truncated branch of depth 3
                      | |--- broad_impact > 355.00
               | | |--- value: [48.00]
                |--- broad impact > 368.00
               | |--- value: [354.00]
              -- publications > 277.00
             |--- broad impact <= 309.50
                | |--- broad impact <= 293.50
                   | |--- broad impact <= 257.50
                       | |--- publications <= 350.50
| | |--- broad_impact <= 218.50
                         | | |--- value: [293.00]
                      | |--- publications > 350.50
| | --- value: [234.00]
                      |--- broad impact > 257.50
                | | | | |--- publications <= 368.50
                | | | | |--- value: [359.00]
                   |--- broad impact > 293.50
                   | |--- publications <= 363.50
                   | | |--- value: [253.00]
               | | |--- publications > 363.50
| | | --- value: [190.00]
                   - broad impact > 309.50
                | |--- publications <= 364.50
                   | |--- publications <= 328.50
                   | | |--- broad_impact <= 342.50
                      | | |--- publications <= 303.50
                              | |--- truncated branch of depth 2
                          | |--- publications > 303.50
                         | | |--- value: [355.00]
                         |--- broad impact > 342.50
                      | | |--- value: [349.00]
                       |--- publications > 328.50
                           |--- publications <= 346.50
                         | |--- value: [380.00]
                         |--- publications > 346.50
                         | |--- value: [395.00]
                      - publications > 364.50
                      |--- publications <= 371.50
                       | |--- value: [275.00]
                       |--- publications > 371.50
```

```
| | | | | |--- value: [353.00]
      |--- publications > 382.00
   | | | |--- value: [40.00]
   |--- publications > 399.00
       |--- broad impact <= 348.00
       | |--- broad impact <= 298.50
          | |--- broad impact <= 258.00
             | |--- broad_impact <= 251.50
              | | |--- value: [321.00]
                  |--- broad_impact > 251.50
              | | |--- value: [317.00]
             |--- broad impact > 258.00
             | |--- publications <= 516.00
              | | |--- value: [329.00]
                  |--- publications > 516.00
             | | |--- value: [340.00]
         |--- broad impact > 298.50
       | | |--- value: [392.00]
       |--- broad impact > 348.00
          |--- broad impact <= 363.50
             |--- value: [425.00]
           |--- broad impact > 363.50
             |--- broad impact <= 372.00
              | |--- publications <= 509.50
              | | |--- value: [456.00]
                  |--- publications > 509.50
                 | |--- value: [461.00]
              |--- broad impact > 372.00
      | | | |--- value: [491.00]
|--- broad impact > 380.50
   |--- broad impact <= 471.50
       |--- publications <= 551.00
       | |--- publications <= 455.50
           | |--- publications <= 321.00
             | |--- publications <= 265.00
              | | |--- value: [418.00]
                 |--- publications > 265.00
| |--- value: [402.00]
              |--- publications > 321.00
              | |--- broad impact <= 427.00
                  | |--- publications <= 436.00
                      | |--- value: [495.00]
                      |--- publications > 436.00
                     | |--- value: [505.00]
                  |--- broad impact > 427.00
                 | |--- publications <= 386.00
                     | |--- publications <= 351.50
                        | |--- value: [484.00]
|--- publications > 351.50
                  | | |--- value: [474.00]
              | | |--- publications > 386.00
              | | | |--- broad_impact <= 433.50
                         | |--- value: [400.00]
                         |--- broad_impact > 433.50
                  | |--- value: [441.00]
                 - publications > 455.50
              |--- publications <= 477.50
              | |--- value: [150.00]
               |--- publications > 477.50
              |--- publications <= 518.50
              | | |--- value: [419.00]
              | |--- publications > 518.50
              | | |--- value: [454.00]
       |--- publications > 551.00
           |--- broad impact <= 430.50
              |--- publications <= 621.00
              | |--- value: [524.00]
              |--- publications > 621.00
              | |--- broad impact <= 416.50
                  | |--- value: [516.00]
                  |--- broad impact > 416.50
              | | |--- value: [519.00]
             -- broad impact > 430.50
             |--- broad impact <= 439.50
              | |--- value: [547.00]
              |--- broad_impact > 439.50
              | |--- value: [541.00]
```

```
| | | |--- broad impact > 471.50
| | | | | |--- value: [114.00]
|--- broad_impact > 477.00
   |--- broad impact <= 709.00
     |--- publications <= 626.50
        |--- broad impact <= 694.50
      | |--- publications <= 592.50
             | |--- publications <= 458.00
                 | |--- broad impact <= 539.00
                     | |--- publications <= 434.50
                        | |--- broad_impact <= 499.50
                           | |--- value: [520.00]
                           |--- broad impact > 499.50
                           | |--- broad impact <= 515.50
                              | |--- value: [562.00]
                        |--- broad impact > 515.50
                     | | |--- value: [563.00]
                     | |--- publications > 434.50
                     | | |--- value: [469.00]
                     |--- broad impact > 539.00
                       |--- publications <= 390.50
                        | |--- value: [401.00]
                     | |--- publications > 390.50
                     | | |--- value: [370.00]
                    -- publications > 458.00
                     |--- publications <= 510.00
                        |--- publications <= 505.50
                         | |--- broad impact <= 652.50
                            | |--- broad impact <= 558.00
                            | | |--- broad_impact <= 525.50
                              | | |--- value: [589.00]
| |--- broad_impact > 525.50
| | |--- value: [608.00]
                               |--- broad impact > 558.00
                              | |--- publications <= 463.50
                            | | | | --- value: [548.00]
                              | |--- publications > 463.50
| | |--- value: [579.00]
                            |--- broad impact > 652.50
                        | | |--- publications <= 488.00
                     | | | | | |--- value: [672.00]
                     |--- publications > 505.50
                     | | |--- value: [422.00]
                     |--- publications > 510.00
                         |--- broad impact <= 638.00
                            |--- broad impact <= 559.50
                            | |--- value: [581.00]
                            |--- broad impact > 559.50
                           | |--- publications <= 567.50
                         | |--- publications > 567.50
                     | | | | |--- value: [678.00]
                    | |--- broau_inpact|
| | |--- value: [745.00]
                        |--- broad impact > 638.00
                 - publications > 592.50
              | |--- publications <= 596.50
                  | |--- broad impact <= 504.50
                    | |--- value: [529.00]
                 | |--- broad_impact > 504.50
                    | |--- value: [258.00]
                  |--- publications > 596.50
                    |--- broad impact <= 573.50
                     | |--- value: [568.00]
                     |--- broad_impact > 573.50
                       |--- broad impact <= 627.50
                        | |--- value: [429.00]
                        |--- broad impact > 627.50
                     | | |--- value: [473.00]
          |--- broad_impact > 694.50
            |--- broad impact <= 701.00
              | |--- value: [729.00]
              |--- broad impact > 701.00
```

```
| | |--- value: [741.00]
    -- publications > 626.50
     |--- broad impact <= 536.00
       |--- publications <= 699.00
        | |--- publications <= 664.00
            | |--- value: [512.00]
        | |--- publications > 664.00
       | | |--- value: [513.00]
       |--- publications > 699.00
           |--- publications <= 754.50
            | |--- value: [585.00]
            |--- publications > 754.50
          | |--- value: [590.00]
        |--- broad impact > 536.00
        |--- broad impact <= 614.00
            |--- publications <= 628.00
            | |--- value: [608.00]
            |--- publications > 628.00
            | |--- broad impact <= 561.50
              | |--- broad_impact <= 553.50
                | | |--- publications <= 740.00
                       | |--- value: [633.00]
                   | | | --- value: [639.00]
                | |--- broad_impact > 553.50
               | | |--- value: [616.00]
                   - broad impact > 561.50
                   |--- publications <= 943.00
                | |--- broad impact <= 582.00
                | | |--- value: [638.00]
                | |--- publications > 662.50
                          | |--- publications <= 686.50
                          | | |--- value: [676.00]
                         | |--- publications > 686.50
                      | | | | | --- truncated branch of depth 2
                      |--- broad_impact > 582.00
                         |--- publications <= 630.50
                          | |--- value: [660.00]
                         |--- publications > 630.50
                      | | | | |--- broad impact <= 605.50
                   | | | | | | --- truncated branch of depth 2
                     | | |--- broad_impact > 605.50
| | | --- value: [670.00]
                   |--- publications > 943.00
               | | |--- value: [623.00]
        |--- broad impact > 614.00
            |--- broad impact <= 661.00
                |--- publications <= 715.00
                | |--- value: [701.00]
                |--- publications > 715.00
                | |--- broad impact <= 639.50
                | | |--- value: [728.00]
                   |--- broad impact > 639.50
                      |--- publications <= 856.00
                      | |--- value: [748.00]
                   | | |--- publications > 856.00
              | | | |--- value: [752.00]
            |--- broad impact > 661.00
               |--- publications <= 780.00
                |--- broad impact <= 669.00
                | | |--- value: [693.00]
                | |--- broad impact > 669.00
                | | |--- value: [762.00]
                |--- publications > 780.00
                   |--- broad impact <= 682.50
                   | |--- value: [607.00]
                   |--- broad impact > 682.50
                   | |--- publications <= 938.50
                       | |--- publications <= 867.00
                         | |--- value: [684.00]
|--- publications > 867.00
                         | |--- value: [719.00]
                       |--- publications > 938.50
                      | |--- value: [644.00]
- broad impact > 709.00
|--- broad_impact <= 915.00
```

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|--- broad impact <= 769.50
   |--- broad impact <= 755.00
       |--- broad impact <= 740.00
           |--- publications <= 803.50
          | |--- publications <= 753.50
| | |--- value: [797.00]
| |--- publications > 753.50
           | | |--- value: [795.00]
          |--- publications > 803.50
           | |--- broad_impact <= 715.50
               | |--- publications <= 838.50
                   | |--- value: [773.00]
               | |--- publications > 838.50
               | | |--- value: [759.00]
           | |--- broad impact > 715.50
                  |--- broad impact <= 727.50
                   | |--- value: [735.00]
                   |--- broad impact > 727.50
         | | | |--- value: [707.00]
      |--- broad impact > 740.00
       | |--- publications <= 686.50
           | |--- value: [776.00]
           |--- publications > 686.50
           | |--- broad impact <= 747.50
           | | |--- value: [808.00]
  |--- broad_impact > 755.00
  | |--- broad impact <= 762.50
       | |--- value: [667.00]
      |--- broad impact > 762.50
  | | |--- value: [639.00]
   - broad impact > 769.50
   |--- publications <= 824.00
       |--- broad impact <= 903.00
        |--- publications <= 651.00
           | |--- broad impact <= 833.50
           | | |--- value: [807.00]
               |--- broad_impact > 833.50
             | |--- value: [756.00]
           |--- publications > 651.00
           | |--- publications <= 760.50
               | |--- broad impact <= 807.50
                     |--- publications <= 750.00
                       | |--- publications <= 732.50
                     | | | --- value: [818.00]
                   | | |--- publications > 732.50
                 | | | |--- value: [819.00]
                  | |--- publications > 750
| | |--- value: [855.00]
                          - publications > 750.00
                  |--- broad impact > 807.50
               | | |--- broad impact <= 870.50
               | | | |--- publications <= 699.00
               | | | | | |--- value: [889.00]
                      | |--- publications > 699.00
| | |--- truncated branch of depth 2
                      |--- broad impact > 870.50
                  | | |--- publications <= 670.50
                  | | | | --- value: [878.00]
                     | |--- publications > 670.50
| | --- value: [834.00]
                  - publications > 760.50
             | |--- publications <= 799.50
                   | |--- broad impact <= 846.50
                   | | |--- value: [804.00]
                   | |--- broad_impact > 846.50
| | --- value: [770.00]
                   |--- publications > 799.50
                   | |--- publications <= 806.50
                   | | |--- value: [843.00]
                  | |--- publications > 806.50
| | |--- value: [833.00]
             i
          -- broad_impact > 903.00
   | | |--- value: [653.00]
   |--- publications > 824.00
   | |--- broad impact <= 868.00
       | |--- publications <= 919.50
```

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|--- broad impact <= 804.50
                   | |--- value: [871.00]
               | |--- broad impact > 804.50
              | | |--- value: [861.00]
           | |--- publications > 919.50
| | |--- value: [892.00]
           |--- broad impact > 868.00
           | |--- broad impact <= 907.50
              | |--- broad impact <= 891.00
          | | | |--- value: [923.00]
           | | |--- broad_impact > 891.00
                   | |--- broad impact <= 896.50
                     | |--- value: [935.00]
                   | |--- broad impact > 896.50
                 | | |--- publications <= 946.50
                   | | | |--- value: [929.00]
               |--- broad impact > 907.50
       | | | |--- value: [858.00]
|--- broad_impact > 915.00
 | |--- broad impact <= 972.50
       |--- broad impact <= 969.50
       | |--- broad impact <= 922.00
        | | |--- publications <= 993.50
       | | | |--- publications <= 898.00
       | | | | |--- value: [920.00]
              | |--- publications > 898.00
| | |--- value: [860.00]
           | |--- publications > 993.50
          | | |--- value: [962.00]
       | |--- broad impact > 922.00
           | |--- publications <= 894.00
                   |--- broad impact <= 951.00
               | | |--- broad_impact <= 934.50
              | | | |--- value: [950.00]
       | | | | |--- broad impact > 934.50
       | | | | | | | --- value: [946.00]
              | |--- broad_impact > 951.00
| | |--- value: [928.00]
              |--- publications > 894.00
       | | | |--- broad impact <= 962.50
 |--- broad impact > 969.50
       | |--- publications <= 936.00
   | | | |--- value: [837.00]
       | |--- publications > 936.00
| | --- value: [807.00]
    |--- broad impact > 972.50
   | |--- publications <= 945.00
       | |--- broad impact <= 978.00
       | | |--- publications <= 873.50
        |--- broad impact > 978.00
          | |--- publications <= 840.00
       | | | |--- value: [985.00]
               |--- publications > 840.00
              | |--- broad_impact <= 985.00
           | | | |--- value: [1000.00]
      | | | | | --- value: [1000.00]
| | | | | --- broad_impact > 985.00
| | | | | | --- value: [994.00]
|--- publications > 945.00
        | |--- value: [920.00]
```

Визуализация дерева

```
return graph.create png()
```

In [37]: Image(get png tree(reg, X.columns), height='100%')

```
Out[37]: 1997

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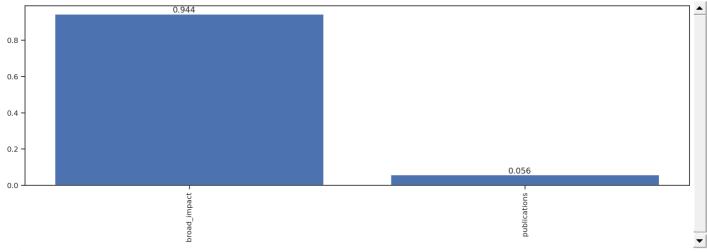
- 1997
```

Важность признаков в дереве

```
In [38]: 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
    \label{list_to_sort} \verb| list_to_sort = list(zip(X_dataset.columns.values, tree_model.feature_importances_))| \\
    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 [39]: boston_tree_regr_fl, boston_tree_regr_fd = draw_feature_importances(reg, X)



In []: