

# MLOps и production подход к ML исследованиям



28 марта - 28 мая



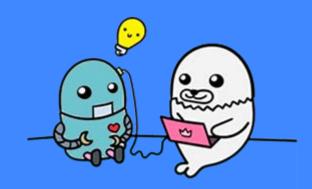


## MLOps и production подход к ML исследованиям

Шаблонизация. Python пакеты и CLI. Snakemake

Павел Кикин

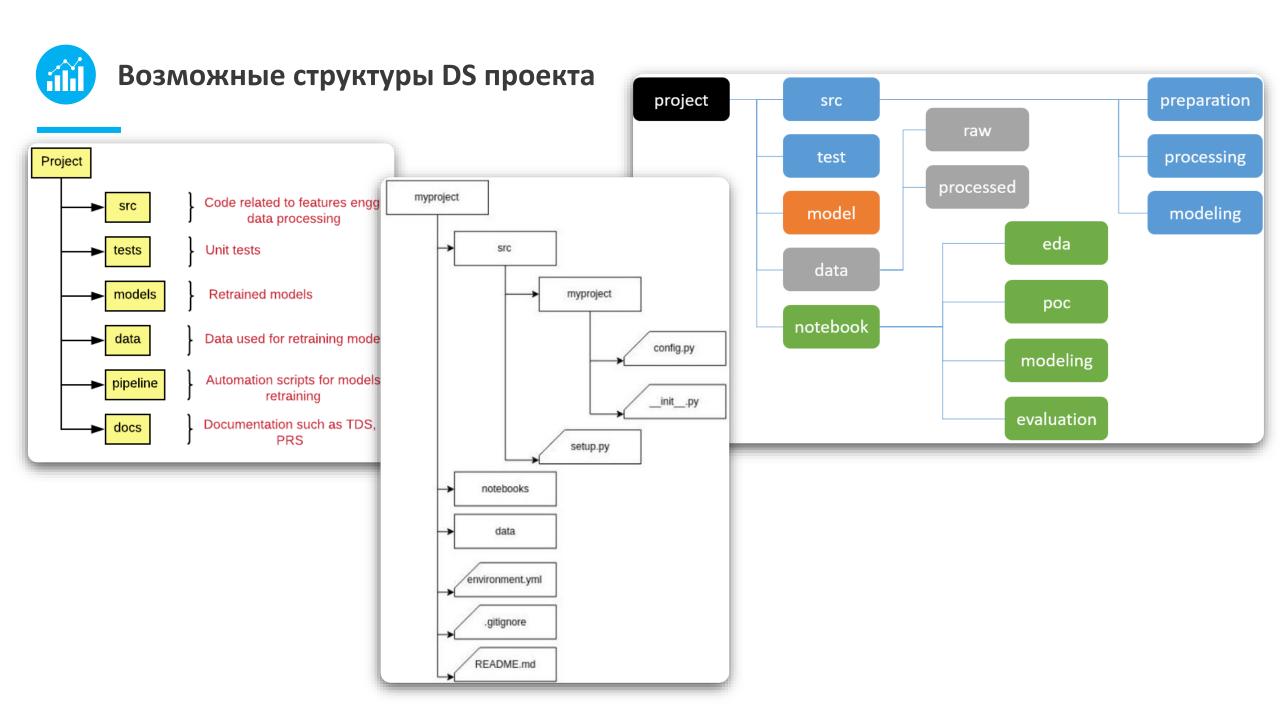
Газпромнефть ЦР Руководитель направления NLP **t.me/pavel\_kikin** 





#### Артефакты DS проекта

- Данные
- EDA, Preprocessing
- Описания, отчеты (MD, Jupyter, latex), графики
- Feature engineering
- Алгоритмы предсказания
- Модели





```
── LICENSE

— Makefile
                      <- Makefile with commands like `make data` or `make train`</p>
- README.md
                      <- The top-level README for developers using this project.
— data
    — external
                      <- Data from third party sources.
    — interim
                      <- Intermediate data that has been transformed.
                      <- The final, canonical data sets for modeling.
    processed
    └─ raw
                      <- The original, immutable data dump.
                      <- A default Sphinx project; see sphinx-doc.org for details
— docs
- models
                      <- Trained and serialized models, model predictions, or model summaries
                      <- Jupyter notebooks. Naming convention is a number (for ordering),
notebooks
                         the creator's initials, and a short `-` delimited description, e.g.
                         `1.0-jqp-initial-data-exploration`.
- references
                      <- Data dictionaries, manuals, and all other explanatory materials.
- reports
                      <- Generated analysis as HTML, PDF, LaTeX, etc.
   └─ figures
                      <- Generated graphics and figures to be used in reporting
requirements.txt <- The requirements file for reproducing the analysis environment, e.g.
                         generated with `pip freeze > requirements.txt`
─ setup.py
                      <- Make this project pip installable with `pip install -e`
— src
                      <- Source code for use in this project.
    ___init__.py <- Makes src a Python module</pre>
     — data
                      <- Scripts to download or generate data
       └─ make dataset.py
    — features
                      <- Scripts to turn raw data into features for modeling
       └── build_features.py
                      <- Scripts to train models and then use trained models to make
    — models
                         predictions
        predict_model.py
       └─ train_model.py
    └── visualization  <- Scripts to create exploratory and results oriented visualizations
       └─ visualize.py
└─ tox.ini
                      <- tox file with settings for running tox; see tox.readthedocs.io
```



#### Создание Cookiecutter шаблона

```
python -m venv ./venv
call ./venv/scripts/activate
```

cookiecutter https://github.com/drivendata/cookiecutter-data-science

```
__all__ = ["echo", "surround", "reverse"]
```

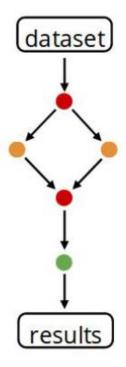
from data.clean\_data import clean\_data
from data.select\_region import select\_region
from features.add\_features import add\_features

- data
- features
- models
- reports
- gitignore ...
- \_\_init\_\_.py

```
mport click
@click.command()
@click.argument("input_path", type=click.Path(exists=True))
@click.argument("output_path", type=click.Path())
@click.argument("region", type=click.INT)
def select_region(input_path: str, output_path: str, region: int):
    """Function selects the listings belonging to a specified region.
    :param input_path: Path to read original DataFrame with all listings
    :param output_path: Path to save filtered DataFrame
   :param region: Selected region id
    :return:
   df = pd.read_csv(input_path)
   df = df[df['region'] == region]
   df.drop('region', axis=1, inplace=True)
   print(f'Selected {len(df)} samples in region {region}.')
   df.to_csv(output_path)
if __name__ == "__main__":
   select_region()
```



#### Workflow менеджеры

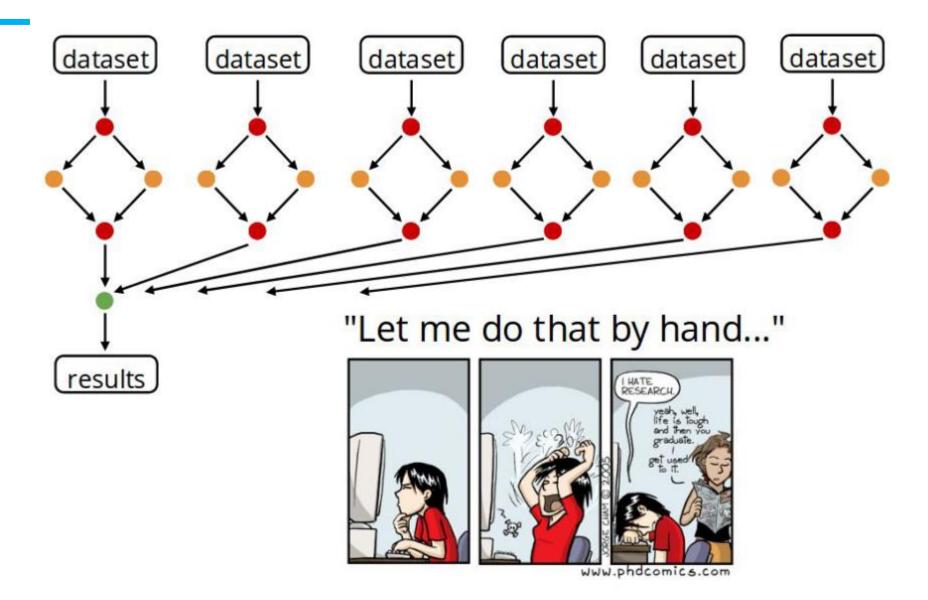


"Let me do that by hand..."

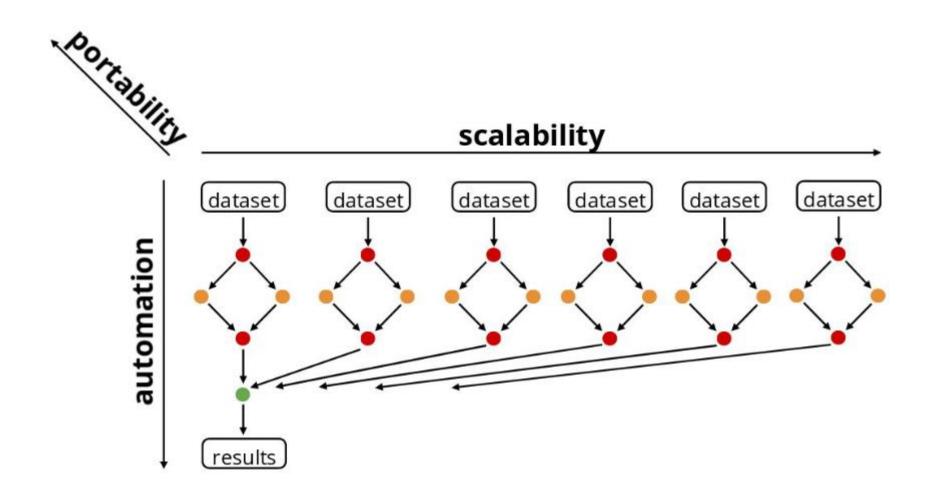


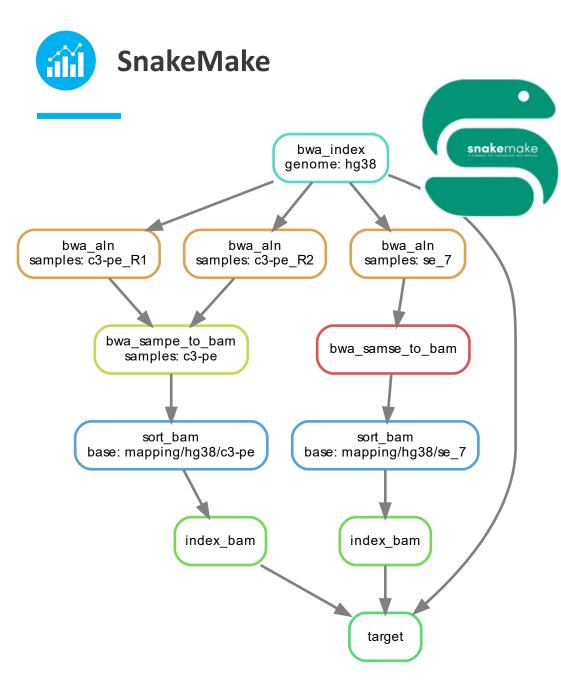


#### Workflow менеджеры



#### Workflow менеджеры





- Декомпозирует анализ на правила, описанные на Python
- Правила определяют как получить выходной файл из входного
- SnakeMake определеяет зависимости и порядок исполнения в виде DAG (направленный ацикличный граф)
- Строит авто-отчеты

# **SnakeMake**

```
# performs the rulegraph generation
snakemake --rulegraph | dot -Tpng >rulegraph.png
# performs the dag generation
snakemake --dag | dot -Tpng >dag.png
# performs dry-run
snakemake -n
# prints shell commands to be executed
snakemake -p
# performs on a subset of rules
snakemake --omit-from rule2 --until rule5
# performs timestamp logs
snakemake --timestamp
```

```
configfile: "workflow/config.yaml'
rule all:
rule add_features:
    shell:
rule clean_data:
    input:
    shell:
rule select_region:
    output:
    shell:
```



- Независимые части DAG могут быть исполнены параллельно
- Максимизирует параллелизм с учетом ресурсов на любое

#### количество:

- Компьютеров
- Кластеров
- Облаков



# execute on cluster
snakemake --cluster qsub --jobs 100
# execute in the cloud

snakemake --kubernetes --jobs 1000 --default-remote-provider GS --default-remote-prefix mybucket

# execute workflow locally with 16 CPU cores

## SnakeMake

Каждое отдельное задание может использовать свое окружение





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