

hmkw5

June 28, 2023

```
[2]: !pip install evidently
```

```
Collecting evidently
  Downloading evidently-0.3.3-py3-none-any.whl (2.0 MB)
    |                               | 2.0 MB 9.5 MB/s
Collecting nltk>=3.6.7
  Downloading nltk-3.8.1-py3-none-any.whl (1.5 MB)
    |                               | 1.5 MB 106.9 MB/s
Requirement already satisfied: pandas>=1.3.5 in
/opt/conda/lib/python3.10/site-packages (from evidently) (2.0.2)
Requirement already satisfied: numpy>=1.19.5 in /opt/conda/lib/python3.10/site-
packages (from evidently) (1.24.3)
Requirement already satisfied: statsmodels>=0.12.2 in
/opt/conda/lib/python3.10/site-packages (from evidently) (0.14.0)
Requirement already satisfied: scipy>=1.5.4 in /opt/conda/lib/python3.10/site-
packages (from evidently) (1.10.1)
Requirement already satisfied: requests>=2.19.0 in
/opt/conda/lib/python3.10/site-packages (from evidently) (2.31.0)
Requirement already satisfied: pydantic<2,>=1.9 in
/opt/conda/lib/python3.10/site-packages (from evidently) (1.10.9)
Collecting plotly>=5.5.0
  Downloading plotly-5.15.0-py2.py3-none-any.whl (15.5 MB)
    |                               | 15.5 MB 107.3 MB/s
Requirement already satisfied: scikit-learn>=0.24.0 in
/opt/conda/lib/python3.10/site-packages (from evidently) (1.2.2)
Requirement already satisfied: PyYAML>=5.1 in /opt/conda/lib/python3.10/site-
packages (from evidently) (6.0)
Collecting umap-learn>=0.5.3
  Downloading umap-learn-0.5.3.tar.gz (88 kB)
    |                               | 88 kB 19.6 MB/s
  Preparing metadata (setup.py) ... done
Requirement already satisfied: joblib in /opt/conda/lib/python3.10/site-
packages (from nltk>=3.6.7->evidently) (1.2.0)
Requirement already satisfied: click in /opt/conda/lib/python3.10/site-packages
(from nltk>=3.6.7->evidently) (8.1.3)
Requirement already satisfied: regex>=2021.8.3 in
/opt/conda/lib/python3.10/site-packages (from nltk>=3.6.7->evidently) (2023.6.3)
Requirement already satisfied: tqdm in /opt/conda/lib/python3.10/site-packages
```

```

(from nltk>=3.6.7->evidently) (4.65.0)
Requirement already satisfied: tzdata>=2022.1 in /opt/conda/lib/python3.10/site-
packages (from pandas>=1.3.5->evidently) (2023.3)
Requirement already satisfied: python-dateutil>=2.8.2 in
/opt/conda/lib/python3.10/site-packages (from pandas>=1.3.5->evidently) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in /opt/conda/lib/python3.10/site-
packages (from pandas>=1.3.5->evidently) (2023.3)
Collecting tenacity>=6.2.0
  Downloading tenacity-8.2.2-py3-none-any.whl (24 kB)
Requirement already satisfied: packaging in /opt/conda/lib/python3.10/site-
packages (from plotly>=5.5.0->evidently) (23.1)
Requirement already satisfied: typing-extensions>=4.2.0 in
/opt/conda/lib/python3.10/site-packages (from pydantic<2,>=1.9->evidently)
(4.6.2)
Requirement already satisfied: charset-normalizer<4,>=2 in
/opt/conda/lib/python3.10/site-packages (from requests>=2.19.0->evidently)
(3.1.0)
Requirement already satisfied: idna<4,>=2.5 in /opt/conda/lib/python3.10/site-
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Requirement already satisfied: certifi>=2017.4.17 in
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(2023.5.7)
Requirement already satisfied: urllib3<3,>=1.21.1 in
/opt/conda/lib/python3.10/site-packages (from requests>=2.19.0->evidently)
(1.26.15)
Requirement already satisfied: threadpoolctl>=2.0.0 in
/opt/conda/lib/python3.10/site-packages (from scikit-learn>=0.24.0->evidently)
(3.1.0)
Requirement already satisfied: patsy>=0.5.2 in /opt/conda/lib/python3.10/site-
packages (from statsmodels>=0.12.2->evidently) (0.5.3)
Requirement already satisfied: numba>=0.49 in /opt/conda/lib/python3.10/site-
packages (from umap-learn>=0.5.3->evidently) (0.57.0)
Collecting pynndescent>=0.5
  Downloading pynndescent-0.5.10.tar.gz (1.1 MB)
    |                               | 1.1 MB 118.1 MB/s
  Preparing metadata (setup.py) ... done
Requirement already satisfied: llvmlite<0.41,>=0.40.0dev0 in
/opt/conda/lib/python3.10/site-packages (from numba>=0.49->umap-
learn>=0.5.3->evidently) (0.40.0)
Requirement already satisfied: six in /opt/conda/lib/python3.10/site-packages
(from patsy>=0.5.2->statsmodels>=0.12.2->evidently) (1.16.0)
Building wheels for collected packages: umap-learn, pynndescent
  Building wheel for umap-learn (setup.py) ... done
  Created wheel for umap-learn: filename=umap_learn-0.5.3-py3-none-any.whl
size=82816
sha256=2a9db83427b3cb37cff3330cf490d19479e41805f7ab3854d1245581e7c5ca5c
  Stored in directory: /home/jovyan/.cache/pip/wheels/a0/e8/c6/a37ea663620bd5200
ea1ba0907ab3c217042c1d035ef606acc

```

```

Building wheel for pynndescent (setup.py) ... done
Created wheel for pynndescent: filename=pynndescent-0.5.10-py3-none-
any.whl size=55622
sha256=29f3ceeed98ee0633187ed6f87e1006c69e6df74e142ef74ca6d1c7813f4303b
Stored in directory: /home/jovyan/.cache/pip/wheels/4a/38/5d/f60a40a66a9512b7e
5e83517ebc2d1b42d857be97d135f1096
Successfully built umap-learn pynndescent
Installing collected packages: tenacity, pynndescent, umap-learn, plotly, nltk,
evidently
Successfully installed evidently-0.3.3 nltk-3.8.1 plotly-5.15.0
pynndescent-0.5.10 tenacity-8.2.2 umap-learn-0.5.3
WARNING: You are using pip version 21.3.1; however, version 23.1.2 is
available.

You should consider upgrading via the '/opt/conda/bin/python3.10 -m pip install
--upgrade pip' command.

```

```

[34]: import requests
import datetime
import pandas as pd

from evidently import ColumnMapping
from evidently.report import Report
from evidently.metrics import ColumnDriftMetric, DatasetDriftMetric,
↳DatasetMissingValuesMetric, ColumnQuantileMetric, RegressionErrorNormality

from joblib import load, dump
from tqdm import tqdm

from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_absolute_error, mean_absolute_percentage_error

```

```

[9]: files = [('green_tripdata_2023-03.parquet', './tripdata'),
↳('green_tripdata_2022-03.parquet', './tripdata')]

print("Download files:")
for file, path in files:
    url=f"https://d37ci6vzurychx.cloudfront.net/trip-data/{file}"
    resp=requests.get(url, stream=True)
    save_path=f"{path}/{file}"
    with open(save_path, "wb") as handle:
        for data in tqdm(resp.iter_content(),
            desc=f"{file}",
            postfix=f"save to {save_path}",
            total=int(resp.headers["Content-Length"])):
            handle.write(data)

```

Download files:

```
green_tripdata_2023-03.parquet: 100%|          | 1730999/1730999 [00:08<00:00,
198205.67it/s, save to ./tripdata/green_tripdata_2023-03.parquet]
green_tripdata_2022-03.parquet: 100%|          | 1615562/1615562 [00:08<00:00,
200619.69it/s, save to ./tripdata/green_tripdata_2022-03.parquet]
```

```
[40]: mar_data = pd.read_parquet('tripdata/green_tripdata_2023-03.parquet')
```

```
[41]: mar_data.describe()
```

```
[41]:
```

	VendorID	lpep_pickup_datetime	\
count	72044.000000	72044	
mean	1.865443	2023-03-16 08:02:42.689842432	
min	1.000000	2023-02-23 16:46:29	
25%	2.000000	2023-03-08 13:55:49.500000	
50%	2.000000	2023-03-16 08:05:27.500000	
75%	2.000000	2023-03-23 21:51:25.500000	
max	2.000000	2023-04-01 00:09:27	
std	0.341252	NaN	

	lpep_dropoff_datetime	RatecodeID	PULocationID	\
count	72044	67507.000000	72044.000000	
mean	2023-03-16 08:20:56.525747968	1.145703	98.437413	
min	2023-02-23 16:55:25	1.000000	1.000000	
25%	2023-03-08 14:15:59.750000128	1.000000	74.000000	
50%	2023-03-16 08:22:27	1.000000	75.000000	
75%	2023-03-23 22:12:31.500000	1.000000	129.000000	
max	2023-04-01 17:32:21	99.000000	265.000000	
std	NaN	1.225735	60.481870	

	DOLocationID	passenger_count	trip_distance	fare_amount	\
count	72044.000000	67507.000000	72044.000000	72044.000000	
mean	137.613556	1.286844	12.449511	17.018203	
min	1.000000	0.000000	0.000000	-115.000000	
25%	74.000000	1.000000	1.100000	9.300000	
50%	138.000000	1.000000	1.840000	13.500000	
75%	216.000000	1.000000	3.300000	20.000000	
max	265.000000	9.000000	92064.680000	477.000000	
std	76.169705	0.923652	641.094653	13.832399	

	extra	mta_tax	tip_amount	tolls_amount	ehail_fee	\
count	72044.000000	72044.000000	72044.000000	72044.000000	0.0	
mean	0.878109	0.576929	2.157651	0.183268	NaN	
min	-5.000000	-0.500000	-1.100000	0.000000	NaN	
25%	0.000000	0.500000	0.000000	0.000000	NaN	
50%	0.000000	0.500000	1.600000	0.000000	NaN	
75%	1.000000	0.500000	3.360000	0.000000	NaN	

max	12.500000	4.250000	270.270000	36.050000	NaN
std	1.297642	0.384129	3.136215	1.139159	NaN

	improvement_surcharge	total_amount	payment_type	trip_type \
count	72044.000000	72044.000000	67507.000000	67500.000000
mean	0.961604	22.292310	1.373680	1.028681
min	-1.000000	-116.000000	1.000000	1.000000
25%	1.000000	12.980000	1.000000	1.000000
50%	1.000000	18.100000	1.000000	1.000000
75%	1.000000	26.730000	2.000000	1.000000
max	1.000000	478.000000	5.000000	2.000000
std	0.185185	15.852047	0.512307	0.166911

	congestion_surcharge
count	67507.000000
mean	0.714837
min	-2.750000
25%	0.000000
50%	0.000000
75%	2.750000
max	2.750000
std	1.206435

```
[42]: # q1 72044
mar_data.shape
```

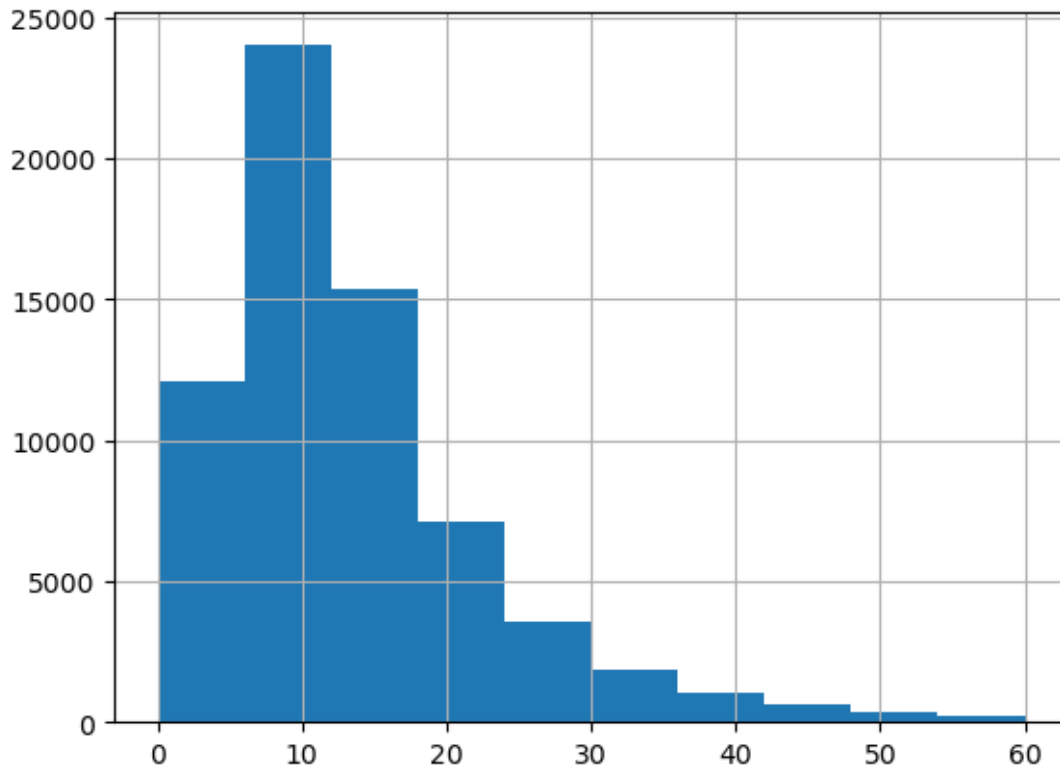
```
[42]: (72044, 20)
```

```
[43]: # create target
mar_data["duration_min"] = mar_data.lpep_dropoff_datetime - mar_data.
    ↳ lpep_pickup_datetime
mar_data.duration_min = mar_data.duration_min.apply(lambda td : float(td.
    ↳ total_seconds())/60)
```

```
[44]: # filter out outliers
mar_data = mar_data[(mar_data.duration_min >= 0) & (mar_data.duration_min <=
    ↳ 60)]
mar_data = mar_data[(mar_data.passenger_count > 0) & (mar_data.passenger_count
    ↳ <= 8)]
```

```
[45]: mar_data.duration_min.hist()
```

```
[45]: <Axes: >
```



```
[46]: # data labeling
target = "duration_min"
num_features = ["passenger_count", "trip_distance", "fare_amount",
               ↪ "total_amount"]
cat_features = ["PULocationID", "DOLocationID"]
```

```
[47]: mar_data.shape
```

```
[47]: (66255, 21)
```

```
[48]: train_data = mar_data[:30000]
      val_data = mar_data[30000:]
```

```
[49]: model = LinearRegression()
```

```
[50]: model.fit(train_data[num_features + cat_features], train_data[target])
```

```
[50]: LinearRegression()
```

```
[51]: train_preds = model.predict(train_data[num_features + cat_features])
      train_data['prediction'] = train_preds
```

```
[52]: val_preds = model.predict(val_data[num_features + cat_features])
      val_data['prediction'] = val_preds
```

```
[53]: print(mean_absolute_error(train_data.duration_min, train_data.prediction))
      print(mean_absolute_error(val_data.duration_min, val_data.prediction))
```

3.5641759571064893

3.646951695051196

Dump model and reference data

```
[54]: with open('models/hmwk5_lin_reg.bin', 'wb') as f_out:
      dump(model, f_out)
```

```
[55]: val_data.to_parquet('tripdata/reference.parquet')
```

Evidently Report

```
[60]: column_mapping = ColumnMapping(
      target='duration_min',
      prediction='prediction',
      numerical_features=num_features,
      categorical_features=cat_features
    )
```

```
[61]: # q2: RegressionErrorNormality
      report = Report(metrics=[
          ColumnDriftMetric(column_name='prediction'),
          DatasetDriftMetric(),
          DatasetMissingValuesMetric(),
          RegressionErrorNormality(),
          ColumnQuantileMetric(column_name='fare_amount', quantile=0.5)
      ])
```

```
[62]: report.run(reference_data=train_data, current_data=val_data,
      ↪column_mapping=column_mapping)
```

```
[63]: report.show(mode='inline')
```

```
[63]: <IPython.core.display.HTML object>
```

```
[64]: result = report.as_dict()
```

```
[65]: result
```

```
[65]: {'metrics': [{'metric': 'ColumnDriftMetric',
      'result': {'column_name': 'prediction',
      'column_type': 'num',
```

```

'stattest_name': 'Wasserstein distance (normed)',
'stattest_threshold': 0.1,
'drift_score': 0.021912744737353492,
'drift_detected': False,
'current': {'small_distribution': {'x': [-29.976938687069985,
    -6.360433663687498,
    17.25607135969499,
    40.87257638307747,
    64.48908140645996,
    88.10558642984245,
    111.72209145322493,
    135.33859647660742,
    158.95510149998992,
    182.57160652337242,
    206.1881115467549],
'y': [9.343432451141208e-06,
    0.03443054858245535,
    0.007515623477886711,
    0.00034570700069222476,
    2.335858112785302e-05,
    8.175503394748558e-06,
    5.839645281963255e-06,
    1.1679290563926504e-06,
    2.335858112785301e-06,
    1.1679290563926519e-06]}}},
'reference': {'small_distribution': {'x': [-44.794580269945996,
    -20.705700872911244,
    3.383178524123508,
    27.472057921158253,
    51.56093731819301,
    75.64981671522777,
    99.7386961122625,
    123.82757550929726,
    147.91645490633204,
    172.0053343033668,
    196.09421370040155],
'y': [1.522140821178265e-05,
    0.00010654985748247855,
    0.03960887169582424,
    0.0016729711389132016,
    7.610704105891324e-05,
    2.4907758892007995e-05,
    2.767528765778663e-06,
    2.767528765778661e-06,
    1.3837643828893314e-06,
    1.3837643828893314e-06]}}}},
{'metric': 'DatasetDriftMetric',

```



```

'result': {'drift_share': 0.5,
'number_of_columns': 8,
'number_of_drifted_columns': 0,
'share_of_drifted_columns': 0.0,
'dataset_drift': False}},
{'metric': 'DatasetMissingValuesMetric',
'result': {'current': {'different_missing_values': {'': 0,
-inf: 0,
None: 36257,
inf: 0},
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-inf: 0,
None: 0,
inf: 0},
'lpep_pickup_datetime': {'': 0, -inf: 0, None: 0, inf: 0},
'lpep_dropoff_datetime': {'': 0, -inf: 0, None: 0, inf: 0},
'store_and_fwd_flag': {'': 0, -inf: 0, None: 0, inf: 0},
'RatecodeID': {'': 0, -inf: 0, None: 0, inf: 0},
'PULocationID': {'': 0, -inf: 0, None: 0, inf: 0},
'DOLocationID': {'': 0, -inf: 0, None: 0, inf: 0},
'passenger_count': {'': 0, -inf: 0, None: 0, inf: 0},
'trip_distance': {'': 0, -inf: 0, None: 0, inf: 0},
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'extra': {'': 0, -inf: 0, None: 0, inf: 0},
'mta_tax': {'': 0, -inf: 0, None: 0, inf: 0},
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'tolls_amount': {'': 0, -inf: 0, None: 0, inf: 0},
'ehail_fee': {'': 0, -inf: 0, None: 36255, inf: 0},
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'payment_type': {'': 0, -inf: 0, None: 0, inf: 0},
'trip_type': {'': 0, -inf: 0, None: 2, inf: 0},
'congestion_surcharge': {'': 0, -inf: 0, None: 0, inf: 0},
'duration_min': {'': 0, -inf: 0, None: 0, inf: 0},
'prediction': {'': 0, -inf: 0, None: 0, inf: 0}}},
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'lpep_pickup_datetime': 0,
'lpep_dropoff_datetime': 0,
'store_and_fwd_flag': 0,
'RatecodeID': 0,
'PULocationID': 0,
'DOLocationID': 0,
'passenger_count': 0,
'trip_distance': 0,
'fare_amount': 0,
'extra': 0,

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```

'mta_tax': 0,
'tip_amount': 0,
'tolls_amount': 0,
'ehail_fee': 1,
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'total_amount': 0,
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'trip_type': 1,
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'lpep_dropoff_datetime': 0,
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'RatecodeID': 0,
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'DOLocationID': 0,
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'trip_distance': 0,
'fare_amount': 0,
'extra': 0,
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'tolls_amount': 0,
'ehail_fee': 36255,
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'lpep_dropoff_datetime': 0.0,
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'RatecodeID': 0.0,
'PULocationID': 0.0,
'DOLocationID': 0.0,
'passenger_count': 0.0,
'trip_distance': 0.0,
'fare_amount': 0.0,
'extra': 0.0,
'mta_tax': 0.0,

```

```

'tip_amount': 0.0,
'tolls_amount': 0.0,
'ehail_fee': 1.0,
'improvement_surcharge': 0.0,
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'payment_type': 0.0,
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'prediction': 0.0},
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'share_of_rows_with_missing_values': 1.0,
'number_of_columns': 22,
'columns_with_missing_values': ['ehail_fee', 'trip_type'],
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'PULocationID': {'': 0, -inf: 0, None: 0, inf: 0},
'DOLocationID': {'': 0, -inf: 0, None: 0, inf: 0},
'passenger_count': {'': 0, -inf: 0, None: 0, inf: 0},
'trip_distance': {'': 0, -inf: 0, None: 0, inf: 0},
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'duration_min': {'': 0, -inf: 0, None: 0, inf: 0},
'prediction': {'': 0, -inf: 0, None: 0, inf: 0}},

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```

'number_of_different_missing_values_by_column': {'VendorID': 0,
'lpep_pickup_datetime': 0,
'lpep_dropoff_datetime': 0,
'store_and_fwd_flag': 0,
'RatecodeID': 0,
'PULocationID': 0,
'DOLocationID': 0,
'passenger_count': 0,
'trip_distance': 0,
'fare_amount': 0,
'extra': 0,
'mta_tax': 0,
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'tolls_amount': 0,
'ehail_fee': 1,
'improvement_surcharge': 0,
'total_amount': 0,
'payment_type': 0,
'trip_type': 0,
'congestion_surcharge': 0,
'duration_min': 0,
'prediction': 0},
'number_of_missing_values': 30000,
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'number_of_missing_values_by_column': {'VendorID': 0,
'lpep_pickup_datetime': 0,
'lpep_dropoff_datetime': 0,
'store_and_fwd_flag': 0,
'RatecodeID': 0,
'PULocationID': 0,
'DOLocationID': 0,
'passenger_count': 0,
'trip_distance': 0,
'fare_amount': 0,
'extra': 0,
'mta_tax': 0,
'tip_amount': 0,
'tolls_amount': 0,
'ehail_fee': 30000,
'improvement_surcharge': 0,
'total_amount': 0,
'payment_type': 0,
'trip_type': 0,
'congestion_surcharge': 0,
'duration_min': 0,
'prediction': 0},
'share_of_missing_values_by_column': {'VendorID': 0.0,

```

```

'lpdp_pickup_datetime': 0.0,
'lpdp_dropoff_datetime': 0.0,
'store_and_fwd_flag': 0.0,
'RatecodeID': 0.0,
'PULocationID': 0.0,
'DOLocationID': 0.0,
'passenger_count': 0.0,
'trip_distance': 0.0,
'fare_amount': 0.0,
'extra': 0.0,
'mta_tax': 0.0,
'tip_amount': 0.0,
'tolls_amount': 0.0,
'ehail_fee': 1.0,
'improvement_surcharge': 0.0,
'total_amount': 0.0,
'payment_type': 0.0,
'trip_type': 0.0,
'congestion_surcharge': 0.0,
'duration_min': 0.0,
'prediction': 0.0},
'number_of_rows': 30000,
'number_of_rows_with_missing_values': 30000,
'share_of_rows_with_missing_values': 1.0,
'number_of_columns': 22,
'columns_with_missing_values': ['ehail_fee'],
'number_of_columns_with_missing_values': 1,
'share_of_columns_with_missing_values': 0.045454545454545456}}},
{'metric': 'RegressionErrorNormality', 'result': {}},
{'metric': 'ColumnQuantileMetric',
 'result': {'column_name': 'fare_amount',
 'column_type': 'num',
 'quantile': 0.5,
 'current': {'value': 12.8},
 'reference': {'value': 12.8}}}}}]

```

```

[26]: #prediction drift
result['metrics'][0]['result']['drift_score']

```

[26]: 0.021912744737353492

```

[27]: #number of drifted columns
result['metrics'][1]['result']['number_of_drifted_columns']

```

[27]: 0

```
[28]: #share of missing values
result['metrics'][2]['result']['current']['share_of_missing_values']
```

[28]: 0.0454570529456752

```
[67]: #share of missing values
print(result['metrics'][4]['result']['current']['value'])
print(result['metrics'][4]['result']['reference']['value'])
```

12.8

12.8

```
[30]: #q3: @task(retries=2, retry_delay_seconds=5, name="calculate metrics")
!pip install psycopg psycopg_binary
```

Collecting psycopg

Downloading psycopg-3.1.9-py3-none-any.whl (167 kB)

| 167 kB 14.9 MB/s

Collecting psycopg_binary

Downloading

psycopg_binary-3.1.9-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl
(3.3 MB)

| 3.3 MB 107.2 MB/s

Requirement already satisfied: typing-extensions>=4.1 in
/opt/conda/lib/python3.10/site-packages (from psycopg) (4.6.2)

Installing collected packages: psycopg-binary, psycopg


Successfully installed psycopg-3.1.9 psycopg-binary-3.1.9

WARNING: You are using pip version 21.3.1; however, version 23.1.2 is
available.


You should consider upgrading via the '/opt/conda/bin/python3.10 -m pip install
--upgrade pip' command.


```
[31]: # open File -> new -> terminal, run the command to have proper access to
      ↪ postgres container
      #!python3 evidently_metrics_calculation.py
```

```
[32]: # add postgresdb connector to grafana (postgresdb - container name, postgres -
      ↪ default user and the password set in docker-compose (disable SSL)
```

 PostgreSQL-1

Type: PostgreSQL

 Settings

 Alerting supported

Name ⓘ

postgresdb

Default ☐

PostgreSQL Connection

Host	postgresdb:5432		
Database	test		
User	postgres	Password
TLS/SSL Mode ⓘ	verify-full ▾		
TLS/SSL Method ⓘ	File system path ▾		

[]: *# check sql query before adding to a dashboard*

PostgreSQL

(PostgreSQL)

Format: Table Filter Group Order Preview

Table

dummy_metrics

Column	Aggregation - optional	Alias - optional	
"timestamp"	Choose	Choose	
quantile_current	MAX	"max_current"	
quantile_reference	MAX	"max_reference"	

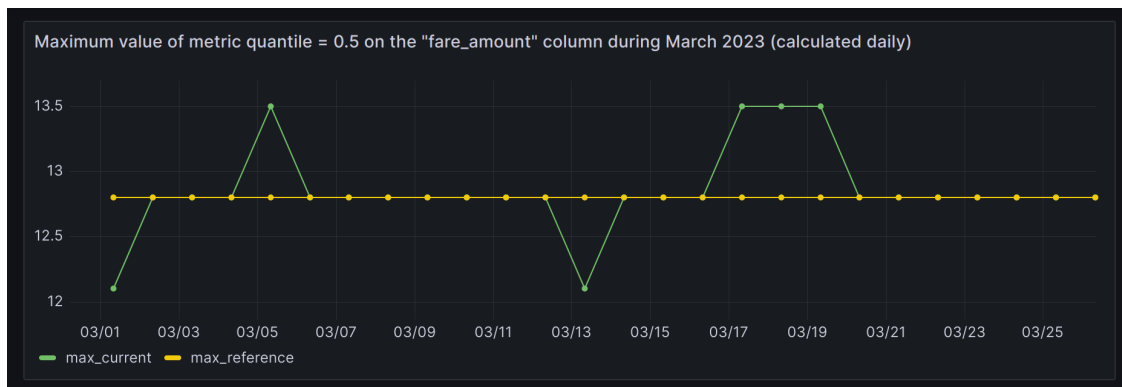
+

Group by column

"timestamp" x +

Preview

```
SELECT
  "timestamp",
  MAX(quantile_current) AS "max_current",
  MAX(quantile_reference) AS "max_reference"
FROM
```



[]: # maximum value of metric quantile = 0.5 on the "fare_amount" column during March 2023 (calculated daily) is 13.5

[]: # json config of the dashboard is saved under project_folder/dashboards