In [1]: !pip install --proxy=http://internet.ford.com:83 pyreadstat

Requirement already satisfied: pyreadstat in /opt/conda/lib/python3.7/site-packages (1.2.0)

Requirement already satisfied: pandas>=1.2.0 in /opt/conda/lib/python3.7/site-package s (from pyreadstat) (1.3.5)

Requirement already satisfied: python-dateutil>=2.7.3 in /opt/conda/lib/python3.7/sit e-packages (from pandas>=1.2.0->pyreadstat) (2.8.2)

Requirement already satisfied: pytz>=2017.3 in /opt/conda/lib/python3.7/site-packages (from pandas>=1.2.0->pyreadstat) (2022.5)

Requirement already satisfied: numpy>=1.17.3 in /opt/conda/lib/python3.7/site-package s (from pandas>=1.2.0->pyreadstat) (1.21.6)

Requirement already satisfied: six>=1.5 in /opt/conda/lib/python3.7/site-packages (fr om python-dateutil>=2.7.3->pandas>=1.2.0->pyreadstat) (1.16.0)

In [2]: !pip install --proxy=http://internet.ford.com:83 gaminet

Requirement already satisfied: gaminet in /opt/conda/lib/python3.7/site-packages (0. 5.8)

Requirement already satisfied: scikit-learn>=0.23.0 in /opt/conda/lib/python3.7/site-packages (from gaminet) (1.0.2)

Requirement already satisfied: matplotlib>=3.1.3 in /opt/conda/lib/python3.7/site-pac kages (from gaminet) (3.5.3)

Requirement already satisfied: tensorflow-lattice>=2.0.8 in /opt/conda/lib/python3.7/ site-packages (from gaminet) (2.0.11)

Requirement already satisfied: numpy>=1.15.2 in /opt/conda/lib/python3.7/site-package s (from gaminet) (1.21.6)

Requirement already satisfied: tensorflow>=2.0.0 in /opt/conda/lib/python3.7/site-pac kages (from gaminet) (2.10.0)

Requirement already satisfied: pandas>=0.19.2 in /opt/conda/lib/python3.7/site-packag es (from gaminet) (1.3.5)

Requirement already satisfied: fonttools>=4.22.0 in /opt/conda/lib/python3.7/site-pac kages (from matplotlib>=3.1.3->gaminet) (4.38.0)

Requirement already satisfied: kiwisolver>=1.0.1 in /opt/conda/lib/python3.7/site-pac kages (from matplotlib>=3.1.3->gaminet) (1.4.4)

Requirement already satisfied: pillow>=6.2.0 in /opt/conda/lib/python3.7/site-package s (from matplotlib>=3.1.3->gaminet) (9.2.0)

Requirement already satisfied: pyparsing>=2.2.1 in /opt/conda/lib/python3.7/site-pack ages (from matplotlib>=3.1.3->gaminet) (3.0.9)

Requirement already satisfied: cycler>=0.10 in /opt/conda/lib/python3.7/site-packages (from matplotlib>=3.1.3->gaminet) (0.11.0)

Requirement already satisfied: python-dateutil>=2.7 in /opt/conda/lib/python3.7/site-packages (from matplotlib>=3.1.3->gaminet) (2.8.2)

Requirement already satisfied: packaging>=20.0 in /opt/conda/lib/python3.7/site-packages (from matplotlib>=3.1.3->gaminet) (21.3)

Requirement already satisfied: pytz>=2017.3 in /opt/conda/lib/python3.7/site-packages (from pandas>=0.19.2->gaminet) (2022.5)

Requirement already satisfied: joblib>=0.11 in /opt/conda/lib/python3.7/site-packages (from scikit-learn>=0.23.0->gaminet) (1.2.0)

Requirement already satisfied: threadpoolctl>=2.0.0 in /opt/conda/lib/python3.7/site-packages (from scikit-learn>=0.23.0->gaminet) (3.1.0)

Requirement already satisfied: scipy>=1.1.0 in /opt/conda/lib/python3.7/site-packages (from scikit-learn>=0.23.0->gaminet) (1.7.3)

Requirement already satisfied: grpcio<2.0,>=1.24.3 in /opt/conda/lib/python3.7/site-p ackages (from tensorflow>=2.0.0->gaminet) (1.50.0)

Requirement already satisfied: tensorflow-estimator<2.11,>=2.10.0 in /opt/conda/lib/p ython3.7/site-packages (from tensorflow>=2.0.0->gaminet) (2.10.0)

Requirement already satisfied: libclang>=13.0.0 in /opt/conda/lib/python3.7/site-pack ages (from tensorflow>=2.0.0->gaminet) (14.0.6)

Requirement already satisfied: wrapt>=1.11.0 in /opt/conda/lib/python3.7/site-package s (from tensorflow>=2.0.0->gaminet) (1.14.1)

Requirement already satisfied: protobuf<3.20,>=3.9.2 in /opt/conda/lib/python3.7/site -packages (from tensorflow>=2.0.0->gaminet) (3.19.6)

Requirement already satisfied: typing-extensions>=3.6.6 in /opt/conda/lib/python3.7/s ite-packages (from tensorflow>=2.0.0->gaminet) (4.4.0)

Requirement already satisfied: flatbuffers>=2.0 in /opt/conda/lib/python3.7/site-pack ages (from tensorflow>=2.0.0->gaminet) (22.10.26)

Requirement already satisfied: h5py>=2.9.0 in /opt/conda/lib/python3.7/site-packages (from tensorflow>=2.0.0->gaminet) (3.7.0)

Requirement already satisfied: keras<2.11,>=2.10.0 in /opt/conda/lib/python3.7/site-p ackages (from tensorflow>=2.0.0->gaminet) (2.10.0)

Requirement already satisfied: google-pasta>=0.1.1 in /opt/conda/lib/python3.7/site-p ackages (from tensorflow>=2.0.0->gaminet) (0.2.0)

Requirement already satisfied: keras-preprocessing>=1.1.1 in /opt/conda/lib/python3.

```
UK COMM (Latest)
7/site-packages (from tensorflow>=2.0.0->gaminet) (1.1.2)
Requirement already satisfied: astunparse>=1.6.0 in /opt/conda/lib/python3.7/site-pac
kages (from tensorflow>=2.0.0->gaminet) (1.6.3)
Requirement already satisfied: six>=1.12.0 in /opt/conda/lib/python3.7/site-packages
(from tensorflow>=2.0.0->gaminet) (1.16.0)
Requirement already satisfied: gast<=0.4.0,>=0.2.1 in /opt/conda/lib/python3.7/site-p
ackages (from tensorflow>=2.0.0->gaminet) (0.4.0)
Requirement already satisfied: tensorboard<2.11,>=2.10 in /opt/conda/lib/python3.7/si
te-packages (from tensorflow>=2.0.0->gaminet) (2.10.1)
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es (from tensorflow>=2.0.0->gaminet) (1.3.0)
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(from tensorflow>=2.0.0->gaminet) (59.8.0)
Requirement already satisfied: tensorflow-io-gcs-filesystem>=0.23.1 in /opt/conda/li
b/python3.7/site-packages (from tensorflow>=2.0.0->gaminet) (0.27.0)
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kages (from tensorflow>=2.0.0->gaminet) (3.3.0)
Requirement already satisfied: graphviz in /opt/conda/lib/python3.7/site-packages (fr
om tensorflow-lattice>=2.0.8->gaminet) (0.20.1)
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rom tensorflow-lattice>=2.0.8->gaminet) (2.0.0)
Requirement already satisfied: wheel<1.0,>=0.23.0 in /opt/conda/lib/python3.7/site-pa
ckages (from astunparse>=1.6.0->tensorflow>=2.0.0->gaminet) (0.37.1)
Requirement already satisfied: requests<3,>=2.21.0 in /opt/conda/lib/python3.7/site-p
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Requirement already satisfied: tensorboard-data-server<0.7.0,>=0.6.0 in /opt/conda/li
b/python3.7/site-packages (from tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet)
(0.6.1)
Requirement already satisfied: markdown>=2.6.8 in /opt/conda/lib/python3.7/site-packa
ges (from tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (3.4.1)
Requirement already satisfied: werkzeug>=1.0.1 in /opt/conda/lib/python3.7/site-packa
ges (from tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (2.2.2)
Requirement already satisfied: google-auth<3,>=1.6.3 in /opt/conda/lib/python3.7/site
-packages (from tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (2.13.0)
Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in /opt/conda/lib/pyt
hon3.7/site-packages (from tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (0.4.
Requirement already satisfied: tensorboard-plugin-wit>=1.6.0 in /opt/conda/lib/python
3.7/site-packages (from tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (1.8.1)
Requirement already satisfied: dm-tree>=0.1.1 in /opt/conda/lib/python3.7/site-packag
es (from dm-sonnet->tensorflow-lattice>=2.0.8->gaminet) (0.1.7)
Requirement already satisfied: tabulate>=0.7.5 in /opt/conda/lib/python3.7/site-packa
ges (from dm-sonnet->tensorflow-lattice>=2.0.8->gaminet) (0.9.0)
Requirement already satisfied: pyasn1-modules>=0.2.1 in /opt/conda/lib/python3.7/site
-packages (from google-auth<3,>=1.6.3->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->ga
minet) (0.2.7)
Requirement already satisfied: cachetools<6.0,>=2.0.0 in /opt/conda/lib/python3.7/sit
e-packages (from google-auth<3,>=1.6.3->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->g
aminet) (5.2.0)
Requirement already satisfied: rsa<5,>=3.1.4 in /opt/conda/lib/python3.7/site-package
s (from google-auth<3,>=1.6.3->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet)
```

Requirement already satisfied: requests-oauthlib>=0.7.0 in /opt/conda/lib/python3.7/s ite-packages (from google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.11,>=2.10->tensorf

low>=2.0.0->gaminet) (1.3.1)

Requirement already satisfied: importlib-metadata>=4.4 in /opt/conda/lib/python3.7/si te-packages (from markdown>=2.6.8->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gamine t) (4.11.4)

Requirement already satisfied: certifi>=2017.4.17 in /opt/conda/lib/python3.7/site-pa ckages (from requests<3,>=2.21.0->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gamine t) (2022.9.24)

Requirement already satisfied: urllib3<1.27,>=1.21.1 in /opt/conda/lib/python3.7/site -packages (from requests<3,>=2.21.0->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gami net) (1.26.11)

Requirement already satisfied: charset-normalizer<3,>=2 in /opt/conda/lib/python3.7/s ite-packages (from requests<3,>=2.21.0->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->g aminet) (2.1.1)

Requirement already satisfied: idna<4,>=2.5 in /opt/conda/lib/python3.7/site-packages (from requests<3,>=2.21.0->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (3.4) Requirement already satisfied: MarkupSafe>=2.1.1 in /opt/conda/lib/python3.7/site-packages (from werkzeug>=1.0.1->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (2.1.1)

Requirement already satisfied: zipp>=0.5 in /opt/conda/lib/python3.7/site-packages (f rom importlib-metadata>=4.4->markdown>=2.6.8->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (3.10.0)

Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in /opt/conda/lib/python3.7/site-packages (from pyasn1-modules>=0.2.1->google-auth<3,>=1.6.3->tensorboard<2.11,>=2.10->tensorflow>=2.0.0->gaminet) (0.4.8)

Requirement already satisfied: oauthlib>=3.0.0 in /opt/conda/lib/python3.7/site-packa ges (from requests-oauthlib>=0.7.0->google-auth-oauthlib<0.5,>=0.4.1->tensorboard<2.1 1,>=2.10->tensorflow>=2.0.0->gaminet) (3.2.2)

In [3]: !pip install --proxy=http://internet.ford.com:83 xlwt

Requirement already satisfied: xlwt in /opt/conda/lib/python3.7/site-packages (1.3.0)

```
In [4]: import pandas as pd
        import numpy as np
        import pyreadstat
        import re
        import os
        import tensorflow as tf
        import json
        from sklearn.preprocessing import MinMaxScaler, OrdinalEncoder
        from sklearn.model_selection import train_test_split
        from sklearn.metrics import r2 score, roc auc score
        from gaminet import GAMINet
        from gaminet.utils import local visualize
        from gaminet.utils import global visualize density
        from gaminet.utils import feature_importance_visualize
        from gaminet.utils import plot trajectory
        from gaminet.utils import plot regularization
```

2022-11-03 22:01:17.448970: I tensorflow/core/platform/cpu feature guard.cc:193] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to us e the following CPU instructions in performance-critical operations: AVX2 FMA To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags. 2022-11-03 22:01:21.566046: E tensorflow/stream executor/cuda/cuda blas.cc:2981] Unab le to register cuBLAS factory: Attempting to register factory for plugin cuBLAS when one has already been registered 2022-11-03 22:01:27.432583: W tensorflow/stream executor/platform/default/dso loader. cc:64] Could not load dynamic library 'libnvinfer.so.7'; dlerror: libnvinfer.so.7: ca nnot open shared object file: No such file or directory; LD LIBRARY PATH: /usr/local/ cuda/lib64:/usr/local/nccl2/lib:/usr/local/cuda/extras/CUPTI/lib64 2022-11-03 22:01:27.432737: W tensorflow/stream executor/platform/default/dso loader. cc:64] Could not load dynamic library 'libnvinfer plugin.so.7'; dlerror: libnvinfer p lugin.so.7: cannot open shared object file: No such file or directory; LD LIBRARY PAT H: /usr/local/cuda/lib64:/usr/local/nccl2/lib:/usr/local/cuda/extras/CUPTI/lib64 2022-11-03 22:01:27.432749: W tensorflow/compiler/tf2tensorrt/utils/py utils.cc:38] T F-TRT Warning: Cannot dlopen some TensorRT libraries. If you would like to use Nvidia GPU with TensorRT, please make sure the missing libraries mentioned above are install ed properly.

```
In [5]: df, meta = pyreadstat.read sas7bdat('uk comm ptrans data.sas7bdat')
        print(df.values.shape)
        (20980, 83)
In [6]: cols = meta.column names
        \# temp df = (df.mean() >= 1)
        # print(temp df.values.shape)
        date cols = [i for i in cols if "date" in i]
        print(date cols)
        # print(df[date_cols].value_counts())
        # print(*cols, sep="\n")
        ['applic_date', 'purchase_date', 'date_into_lru']
In [7]: def cat or num(col):
            if(df["odefind_new"].value_counts().shape[0] == 2):
                 return {"type":"categorical"}
            else:
                 return {"type":"continuous"}
        meta info = {}
        col odefind1 = None
        for col in sorted(cols):
            if(col != "odefind1" and col not in date_cols):
                meta info[col] = cat or num(col)
            else:
                try:
                     req = df.pop(col)
                     if(col == "odefind1"):
                         df["odefind1"] = req
                except Exception:
                     print("Already popped")
        meta info["odefind1"] = {"type":"target"}
        print(meta info)
        print(len(meta_info))
```

{'acidtest_1': {'type': 'categorical'}, 'advance_pct': {'type': 'categorical'}, 'avgt

```
erm': {'type': 'categorical'}, 'balloon_amt': {'type': 'categorical'}, 'balloon_ind':
        {'type': 'categorical'}, 'business': {'type': 'categorical'}, 'capitalemployed_1':
        {'type': 'categorical'}, 'ccjtoshftotl24m_1': {'type': 'categorical'}, 'compage_1':
        {'type': 'categorical'}, 'compavgdbt_12mth_1': {'type': 'categorical'}, 'compavgdbt_3
        mth_1': {'type': 'categorical'}, 'compavgdbt_6mth_1': {'type': 'categorical'}, 'compd
        bt_1': {'type': 'categorical'}, 'compnumdbt_100000_1': {'type': 'categorical'}, 'comp
        numdbt_10000_1': {'type': 'categorical'}, 'compunpaccs_1': {'type': 'categorical'},
        'country': {'type': 'categorical'}, 'creditlimit_1': {'type': 'categorical'}, 'curren
        tliabilitiestotal_1': {'type': 'categorical'}, 'curtoprevdirlst12mth_1': {'type': 'ca
        tegorical'}, 'dl12_regnumber_1': {'type': 'categorical'}, 'dl41_supppaypattn_1': {'ty
        pe': 'categorical'}, 'dl77_creditrating_1': {'type': 'categorical'}, 'dl80_region_1':
        {'type': 'categorical'}, 'dl80_siccode1992_1': {'type': 'categorical'}, 'equitygear_
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        {'type': 'categorical'}, 'indavgdbt_3mth_1': {'type': 'categorical'}, 'indavgdbt_6mth
        _1': {'type': 'categorical'}, 'inddbt_1': {'type': 'categorical'}, 'indexpmth_1': {'t
        ype': 'categorical'}, 'lru_mth': {'type': 'categorical'}, 'lru_qtr': {'type': 'catego
        rical'}, 'numccjlast12_1': {'type': 'categorical'}, 'numccls_1': {'type': 'categorica
        l'}, 'numcurdirs_1': {'type': 'categorical'}, 'numdirwshares_1': {'type': 'categorica
        1'}, 'numprevsrchs_12mths_1': {'type': 'categorical'}, 'numprevsrchs_3mths_1': {'type'
        e': 'categorical'}, 'numprevsrchs_6mths_1': {'type': 'categorical'}, 'numsatmortchrge
        s_1': {'type': 'categorical'}, 'numunpaidacctslstmth_1': {'type': 'categorical'}, 'nu
        myrsacctsavail_1': {'type': 'categorical'}, 'numyrspretaxproflst4yrs_1': {'type': 'ca
        tegorical'}, 'odefind_new': {'type': 'categorical'}, 'percapptslst12mth_1': {'type':
        'categorical'}, 'perccapemploftotassts_1': {'type': 'categorical'}, 'percchngecapempl
        d_1': {'type': 'categorical'}, 'percchngecurratio_1': {'type': 'categorical'}, 'percc
        hngeshfunds_1': {'type': 'categorical'}, 'percchngetotassts_1': {'type': 'categorica
        l'}, 'perccurliaboftotassts_1': {'type': 'categorical'}, 'percfasstsoftotassts_1':
        {'type': 'categorical'}, 'percstockofcurassts_1': {'type': 'categorical'}, 'phwt':
        {'type': 'categorical'}, 'phwtaa': {'type': 'categorical'}, 'phwtdf': {'type': 'categ
        orical'}, 'ppavcurrdbt_1': {'type': 'categorical'}, 'product': {'type': 'categorica
        1'}, 'product2': {'type': 'categorical'}, 'purch_mth': {'type': 'categorical'}, 'purc
        h_qtr': {'type': 'categorical'}, 'riskscore_1': {'type': 'categorical'}, 'sco1_pop_fi
        nal': {'type': 'categorical'}, 'scope_refno': {'type': 'categorical'}, 'shorttermloan
        s_1': {'type': 'categorical'}, 'sic92_2': {'type': 'categorical'}, 'sincelastacctssub
        mtd_1': {'type': 'categorical'}, 'sincelastretnssubmtd_1': {'type': 'categorical'},
        'slife_km': {'type': 'categorical'}, 'term': {'type': 'categorical'}, 'term_ds': {'ty
        pe': 'categorical'}, 'time_in_business': {'type': 'categorical'}, 'totallongtermliabi
        lities_1': {'type': 'categorical'}, 'totassets_1': {'type': 'categorical'}, 'vehicle_
        age': {'type': 'categorical'}, 'odefind1': {'type': 'target'}}
In [8]:
        nan_cols = pd.DataFrame(df.isna().sum())
        nan100_cols = pd.DataFrame(df.isna().sum()*100/df.values.shape[0])
        print(nan100_cols)
```

```
In [8]: nan_cols = pd.DataFrame(df.isna().sum())
    nan100_cols = pd.DataFrame(df.isna().sum()*100/df.values.shape[0])
    print(nan100_cols)
    nan_df = pd.concat([nan_cols, nan100_cols], axis=1, join='inner')
    print(nan_df)
    nan_df.to_excel("information.xls", sheet_name="Nan_Columns")
```

```
0
scope_refno
                         0.000000
                         0.000000
advance_pct
balloon amt
                         0.000000
final term
                         0.000000
time_in_business
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percapptslst12mth_1
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numccls 1
                         0.934223
compunpaccs 1
                        28.517636
numunpaidacctslstmth 1
                         3.112488
odefind1
                         0.000000
[80 rows x 1 columns]
scope refno
                           0
                               0.000000
advance pct
                               0.000000
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balloon amt
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final term
                               0.000000
time_in_business
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                         . . .
percapptslst12mth 1
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                         196
numccls 1
                               0.934223
                        5983 28.517636
compunpaccs 1
numunpaidacctslstmth_1
                         653
                               3.112488
odefind1
                               0.000000
```

[80 rows x 2 columns]

/opt/conda/lib/python3.7/site-packages/ipykernel_launcher.py:6: FutureWarning: As the xlwt package is no longer maintained, the xlwt engine will be removed in a future ver sion of pandas. This is the only engine in pandas that supports writing in the xls fo rmat. Install openpyxl and write to an xlsx file instead. You can set the option io.e xcel.xls.writer to 'xlwt' to silence this warning. While this option is deprecated an d will also raise a warning, it can be globally set and the warning suppressed.

Imputing and Removing specified colums:

```
1. Remove compnumdbt_100000_1 and scope_refno
```

2. Imputing remaining

```
else:
                 df[key].fillna(df[key].mean(), inplace=True)
In [11]: x, y = df.iloc[:,:-1].values, df.iloc[:,[-1]].values
         print(x.shape, y.shape)
         xx = np.zeros((x.shape[0], x.shape[1]), dtype=np.float32)
         for i, (key, item) in enumerate(meta_info.items()):
             if item['type'] == 'target':
                 sy = MinMaxScaler((0,1))
                 y = sy.fit_transform(y)
                 meta info[key]['scaler'] = sy
             elif item['type'] == 'categorical':
                 enc = OrdinalEncoder()
                 xx[:,[i]] = enc.fit_transform(x[:,[i]])
                 meta_info[key]['values'] = []
                 for item in enc.categories [0].tolist():
                     try:
                         if item == int(item):
                             meta_info[key]['values'].append(str(int(item)))
                         else:
                             meta info[key]['values'].append(str(item))
                     except ValueError:
                         meta info[key]['values'].append(str(item))
             else:
                 sx = MinMaxScaler((0, 1))
                 xx[:,[i]] = sx.fit_transform(x[:,[i]])
                 meta_info[key]['scaler'] = sx
         task type = 'Regression'
         train_x, test_x, train_y, test_y = train_test_split(xx.astype(np.float32), y, test_siz
         print(train_x.shape, train_y.shape)
         (20980, 77) (20980, 1)
         (14686, 77) (14686, 1)
In [12]: def metric_wrapper(metric, scaler):
             def wrapper(label, pred):
                 return metric(label, pred, scaler=scaler)
             return wrapper
         def ras(label, pred, scaler=None):
             return r2_score(label, pred)
         get_metric = metric_wrapper(ras, None)
         ______
In [13]: model = GAMINet(meta info=meta info, interact num=30,
                         interact_arch=[100] * 5, subnet_arch=[100] * 5,
                         batch_size=100, task_type=task_type, activation_func=tf.nn.relu,
                         main effect epochs=50, interaction epochs=50, tuning epochs=100,
                         lr bp=[0.0001, 0.0001, 0.0001], early stop thres=[50, 50, 50],
                         heredity=True, loss_threshold=0.01, reg_clarity=0.1,
                         mono_increasing_list=[], mono_decreasing_list=[], ## the indices list
                         verbose=True, val_ratio=0.2, random_state=0)
```

model.fit(train_x, train_y)

2022-11-03 22:01:38.287403: I tensorflow/stream_executor/cuda/cuda_gpu_executor.cc:98 0] successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero

2022-11-03 22:01:38.335063: W tensorflow/stream_executor/platform/default/dso_loader. cc:64] Could not load dynamic library 'libcusolver.so.11'; dlerror: libcusolver.so.1 1: cannot open shared object file: No such file or directory; LD_LIBRARY_PATH: /usr/local/cuda/lib64:/usr/local/nccl2/lib:/usr/local/cuda/extras/CUPTI/lib64

2022-11-03 22:01:38.343833: W tensorflow/core/common_runtime/gpu/gpu_device.cc:1934] Cannot dlopen some GPU libraries. Please make sure the missing libraries mentioned ab ove are installed properly if you would like to use GPU. Follow the guide at https://www.tensorflow.org/install/gpu for how to download and setup the required libraries f or your platform.

Skipping registering GPU devices...

2022-11-03 22:01:38.374650: I tensorflow/core/platform/cpu_feature_guard.cc:193] This TensorFlow binary is optimized with oneAPI Deep Neural Network Library (oneDNN) to us e the following CPU instructions in performance-critical operations: AVX2 FMA To enable them in other operations, rebuild TensorFlow with the appropriate compiler flags.

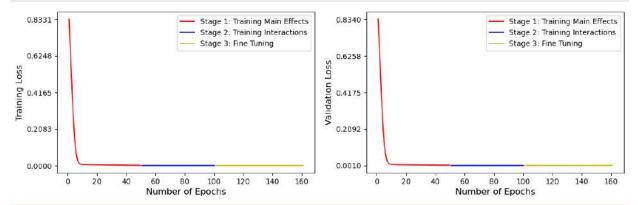
```
########Stage 1: main effect training start.########
Main effects training epoch: 1, train loss: 0.83310, val loss: 0.83401
Main effects training epoch: 2, train loss: 0.63944, val loss: 0.64192
Main effects training epoch: 3, train loss: 0.44092, val loss: 0.44475
Main effects training epoch: 4, train loss: 0.26800, val loss: 0.27261
Main effects training epoch: 5, train loss: 0.14116, val loss: 0.14575
Main effects training epoch: 6, train loss: 0.06432, val loss: 0.06823
Main effects training epoch: 7, train loss: 0.02665, val loss: 0.02959
Main effects training epoch: 8, train loss: 0.01196, val loss: 0.01404
Main effects training epoch: 9, train loss: 0.00724, val loss: 0.00871
Main effects training epoch: 10, train loss: 0.00580, val loss: 0.00694
Main effects training epoch: 11, train loss: 0.00525, val loss: 0.00623
Main effects training epoch: 12, train loss: 0.00493, val loss: 0.00584
Main effects training epoch: 13, train loss: 0.00467, val loss: 0.00556
Main effects training epoch: 14, train loss: 0.00446, val loss: 0.00535
Main effects training epoch: 15, train loss: 0.00429, val loss: 0.00517
Main effects training epoch: 16, train loss: 0.00413, val loss: 0.00502
Main effects training epoch: 17, train loss: 0.00399, val loss: 0.00489
Main effects training epoch: 18, train loss: 0.00386, val loss: 0.00477
Main effects training epoch: 19, train loss: 0.00373, val loss: 0.00466
Main effects training epoch: 20, train loss: 0.00362, val loss: 0.00456
Main effects training epoch: 21, train loss: 0.00351, val loss: 0.00447
Main effects training epoch: 22, train loss: 0.00340, val loss: 0.00437
Main effects training epoch: 23, train loss: 0.00330, val loss: 0.00429
Main effects training epoch: 24, train loss: 0.00319, val loss: 0.00421
Main effects training epoch: 25, train loss: 0.00309, val loss: 0.00412
Main effects training epoch: 26, train loss: 0.00299, val loss: 0.00403
Main effects training epoch: 27, train loss: 0.00289, val loss: 0.00396
Main effects training epoch: 28, train loss: 0.00279, val loss: 0.00388
Main effects training epoch: 29, train loss: 0.00270, val loss: 0.00380
Main effects training epoch: 30, train loss: 0.00260, val loss: 0.00373
Main effects training epoch: 31, train loss: 0.00251, val loss: 0.00365
Main effects training epoch: 32, train loss: 0.00241, val loss: 0.00358
Main effects training epoch: 33, train loss: 0.00232, val loss: 0.00351
Main effects training epoch: 34, train loss: 0.00222, val loss: 0.00344
Main effects training epoch: 35, train loss: 0.00214, val loss: 0.00338
Main effects training epoch: 36, train loss: 0.00204, val loss: 0.00330
Main effects training epoch: 37, train loss: 0.00195, val loss: 0.00322
Main effects training epoch: 38, train loss: 0.00185, val loss: 0.00314
Main effects training epoch: 39, train loss: 0.00176, val loss: 0.00311
Main effects training epoch: 40, train loss: 0.00167, val loss: 0.00300
Main effects training epoch: 41, train loss: 0.00158, val loss: 0.00294
Main effects training epoch: 42, train loss: 0.00149, val loss: 0.00289
Main effects training epoch: 43, train loss: 0.00141, val loss: 0.00282
Main effects training epoch: 44, train loss: 0.00133, val loss: 0.00277
Main effects training epoch: 45, train loss: 0.00124, val loss: 0.00271
Main effects training epoch: 46, train loss: 0.00116, val loss: 0.00266
Main effects training epoch: 47, train loss: 0.00109, val loss: 0.00257
Main effects training epoch: 48, train loss: 0.00101, val loss: 0.00255
Main effects training epoch: 49, train loss: 0.00093, val loss: 0.00250
Main effects training epoch: 50, train loss: 0.00086, val loss: 0.00244
########Stage 1: main effect training stop.########
########Stage 2: interaction training start.########
Interaction training epoch: 1, train loss: 0.00007, val loss: 0.00113
Interaction training epoch: 2, train loss: 0.00005, val loss: 0.00105
Interaction training epoch: 3, train loss: 0.00004, val loss: 0.00106
```

```
Interaction training epoch: 4, train loss: 0.00003, val loss: 0.00102
Interaction training epoch: 5, train loss: 0.00002, val loss: 0.00098
Interaction training epoch: 6, train loss: 0.00002, val loss: 0.00103
Interaction training epoch: 7, train loss: 0.00002, val loss: 0.00105
Interaction training epoch: 8, train loss: 0.00002, val loss: 0.00104
Interaction training epoch: 9, train loss: 0.00001, val loss: 0.00106
Interaction training epoch: 10, train loss: 0.00002, val loss: 0.00100
Interaction training epoch: 11, train loss: 0.00001, val loss: 0.00102
Interaction training epoch: 12, train loss: 0.00001, val loss: 0.00104
Interaction training epoch: 13, train loss: 0.00002, val loss: 0.00111
Interaction training epoch: 14, train loss: 0.00001, val loss: 0.00107
Interaction training epoch: 15, train loss: 0.00001, val loss: 0.00105
Interaction training epoch: 16, train loss: 0.00001, val loss: 0.00108
Interaction training epoch: 17, train loss: 0.00001, val loss: 0.00106
Interaction training epoch: 18, train loss: 0.00001, val loss: 0.00109
Interaction training epoch: 19, train loss: 0.00001, val loss: 0.00099
Interaction training epoch: 20, train loss: 0.00001, val loss: 0.00110
Interaction training epoch: 21, train loss: 0.00001, val loss: 0.00098
Interaction training epoch: 22, train loss: 0.00001, val loss: 0.00113
Interaction training epoch: 23, train loss: 0.00001, val loss: 0.00102
Interaction training epoch: 24, train loss: 0.00001, val loss: 0.00103
Interaction training epoch: 25, train loss: 0.00001, val loss: 0.00101
Interaction training epoch: 26, train loss: 0.00001, val loss: 0.00102
Interaction training epoch: 27, train loss: 0.00001, val loss: 0.00102
Interaction training epoch: 28, train loss: 0.00001, val loss: 0.00101
Interaction training epoch: 29, train loss: 0.00001, val loss: 0.00109
Interaction training epoch: 30, train loss: 0.00001, val loss: 0.00103
Interaction training epoch: 31, train loss: 0.00001, val loss: 0.00101
Interaction training epoch: 32, train loss: 0.00001, val loss: 0.00104
Interaction training epoch: 33, train loss: 0.00001, val loss: 0.00106
Interaction training epoch: 34, train loss: 0.00001, val loss: 0.00107
Interaction training epoch: 35, train loss: 0.00001, val loss: 0.00103
Interaction training epoch: 36, train loss: 0.00001, val loss: 0.00108
Interaction training epoch: 37, train loss: 0.00001, val loss: 0.00104
Interaction training epoch: 38, train loss: 0.00001, val loss: 0.00105
Interaction training epoch: 39, train loss: 0.00001, val loss: 0.00103
Interaction training epoch: 40, train loss: 0.00001, val loss: 0.00102
Interaction training epoch: 41, train loss: 0.00001, val loss: 0.00103
Interaction training epoch: 42, train loss: 0.00001, val loss: 0.00103
Interaction training epoch: 43, train loss: 0.00001, val loss: 0.00103
Interaction training epoch: 44, train loss: 0.00001, val loss: 0.00103
Interaction training epoch: 45, train loss: 0.00001, val loss: 0.00109
Interaction training epoch: 46, train loss: 0.00000, val loss: 0.00104
Interaction training epoch: 47, train loss: 0.00001, val loss: 0.00109
Interaction training epoch: 48, train loss: 0.00001, val loss: 0.00106
Interaction training epoch: 49, train loss: 0.00001, val loss: 0.00100
Interaction training epoch: 50, train loss: 0.00001, val loss: 0.00103
########Stage 2: interaction training stop.#########
Fine tuning epoch: 1, train loss: 0.00001, val loss: 0.00109
Fine tuning epoch: 2, train loss: 0.00001, val loss: 0.00104
Fine tuning epoch: 3, train loss: 0.00001, val loss: 0.00102
Fine tuning epoch: 4, train loss: 0.00001, val loss: 0.00108
Fine tuning epoch: 5, train loss: 0.00000, val loss: 0.00103
Fine tuning epoch: 6, train loss: 0.00001, val loss: 0.00109
Fine tuning epoch: 7, train loss: 0.00000, val loss: 0.00103
Fine tuning epoch: 8, train loss: 0.00001, val loss: 0.00108
Fine tuning epoch: 9, train loss: 0.00000, val loss: 0.00107
```

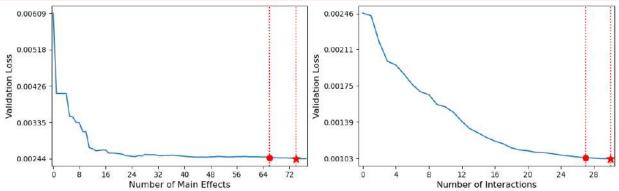
```
Fine tuning epoch: 10, train loss: 0.00000, val loss: 0.00101
Fine tuning epoch: 11, train loss: 0.00001, val loss: 0.00102
Fine tuning epoch: 12, train loss: 0.00000, val loss: 0.00105
Fine tuning epoch: 13, train loss: 0.00000, val loss: 0.00107
Fine tuning epoch: 14, train loss: 0.00000, val loss: 0.00104
Fine tuning epoch: 15, train loss: 0.00000, val loss: 0.00105
Fine tuning epoch: 16, train loss: 0.00001, val loss: 0.00102
Fine tuning epoch: 17, train loss: 0.00000, val loss: 0.00106
Fine tuning epoch: 18, train loss: 0.00000, val loss: 0.00106
Fine tuning epoch: 19, train loss: 0.00000, val loss: 0.00106
Fine tuning epoch: 20, train loss: 0.00001, val loss: 0.00111
Fine tuning epoch: 21, train loss: 0.00000, val loss: 0.00105
Fine tuning epoch: 22, train loss: 0.00000, val loss: 0.00102
Fine tuning epoch: 23, train loss: 0.00000, val loss: 0.00110
Fine tuning epoch: 24, train loss: 0.00000, val loss: 0.00107
Fine tuning epoch: 25, train loss: 0.00000, val loss: 0.00108
Fine tuning epoch: 26, train loss: 0.00000, val loss: 0.00106
Fine tuning epoch: 27, train loss: 0.00000, val loss: 0.00106
Fine tuning epoch: 28, train loss: 0.00000, val loss: 0.00107
Fine tuning epoch: 29, train loss: 0.00000, val loss: 0.00106
Fine tuning epoch: 30, train loss: 0.00000, val loss: 0.00111
Fine tuning epoch: 31, train loss: 0.00000, val loss: 0.00109
Fine tuning epoch: 32, train loss: 0.00001, val loss: 0.00117
Fine tuning epoch: 33, train loss: 0.00001, val loss: 0.00116
Fine tuning epoch: 34, train loss: 0.00000, val loss: 0.00113
Fine tuning epoch: 35, train loss: 0.00000, val loss: 0.00110
Fine tuning epoch: 36, train loss: 0.00001, val loss: 0.00113
Fine tuning epoch: 37, train loss: 0.00000, val loss: 0.00115
Fine tuning epoch: 38, train loss: 0.00000, val loss: 0.00116
Fine tuning epoch: 39, train loss: 0.00000, val loss: 0.00112
Fine tuning epoch: 40, train loss: 0.00000, val loss: 0.00118
Fine tuning epoch: 41, train loss: 0.00000, val loss: 0.00117
Fine tuning epoch: 42, train loss: 0.00000, val loss: 0.00118
Fine tuning epoch: 43, train loss: 0.00000, val loss: 0.00119
Fine tuning epoch: 44, train loss: 0.00000, val loss: 0.00122
Fine tuning epoch: 45, train loss: 0.00001, val loss: 0.00116
Fine tuning epoch: 46, train loss: 0.00001, val loss: 0.00126
Fine tuning epoch: 47, train loss: 0.00001, val loss: 0.00124
Fine tuning epoch: 48, train loss: 0.00001, val loss: 0.00128
Fine tuning epoch: 49, train loss: 0.00001, val loss: 0.00129
Fine tuning epoch: 50, train loss: 0.00001, val loss: 0.00122
Fine tuning epoch: 51, train loss: 0.00001, val loss: 0.00136
Fine tuning epoch: 52, train loss: 0.00000, val loss: 0.00131
Fine tuning epoch: 53, train loss: 0.00000, val loss: 0.00133
Fine tuning epoch: 54, train loss: 0.00000, val loss: 0.00131
Fine tuning epoch: 55, train loss: 0.00001, val loss: 0.00138
Fine tuning epoch: 56, train loss: 0.00001, val loss: 0.00138
Fine tuning epoch: 57, train loss: 0.00001, val loss: 0.00143
Fine tuning epoch: 58, train loss: 0.00001, val loss: 0.00144
Fine tuning epoch: 59, train loss: 0.00001, val loss: 0.00142
Fine tuning epoch: 60, train loss: 0.00001, val loss: 0.00143
Fine tuning epoch: 61, train loss: 0.00001, val loss: 0.00147
Early stop at epoch 61, with validation loss: 0.00147
```

```
In [14]: folder = "./imgs/"
   if not os.path.exists(folder):
```

os.makedirs(folder)
data_dict_logs = model.summary_logs(save_dict=False)
plot_trajectory(data_dict_logs, folder=folder, name="DATASET_TRAJ", save_png=True, sav
plot_regularization(data_dict_logs, folder=folder, name="DATASET_REGU", save_png=True,



The PostScript backend does not support transparency; partially transparent artists will be rendered opaque.



In [15]: data_dict_global = model.global_explain(save_dict=True, folder=folder, name="gaminet_g
global_visualize_density(data_dict_global, folder=folder, name="gaminet_global", main_

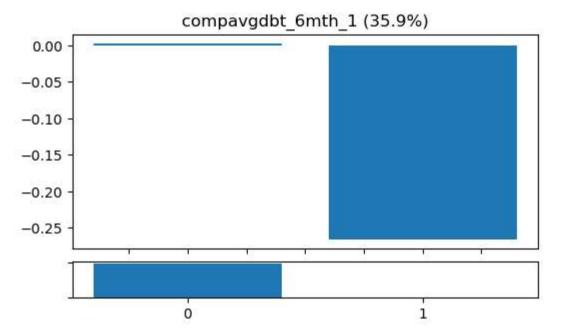
2022-11-03 23:22:01.638956: W tensorflow/core/framework/cpu_allocator_impl.cc:82] All ocation of 1914072192 exceeds 10% of free system memory.

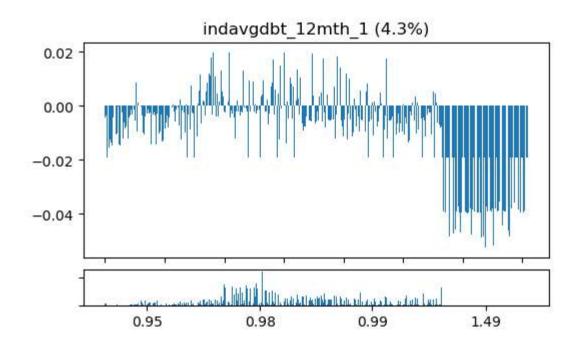
2022-11-03 23:22:06.329700: W tensorflow/core/framework/cpu_allocator_impl.cc:82] All ocation of 1914319680 exceeds 10% of free system memory.

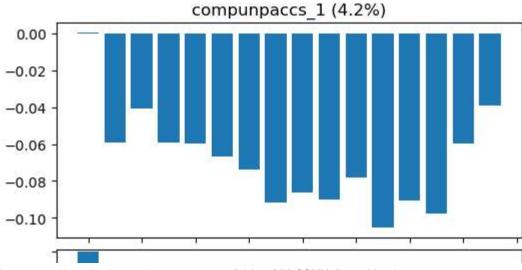
2022-11-03 23:22:11.565525: W tensorflow/core/framework/cpu_allocator_impl.cc:82] All ocation of 1977701832 exceeds 10% of free system memory.

2022-11-03 23:22:15.982688: W tensorflow/core/framework/cpu_allocator_impl.cc:82] All ocation of 1977953400 exceeds 10% of free system memory.

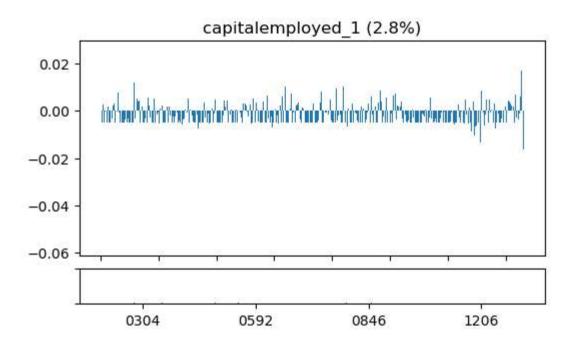
2022-11-03 23:22:21.832548: W tensorflow/core/framework/cpu_allocator_impl.cc:82] All ocation of 1902951432 exceeds 10% of free system memory.

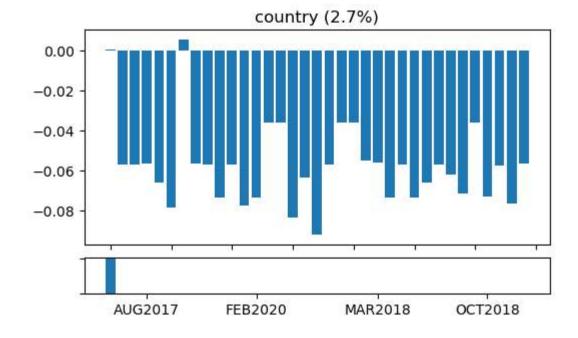


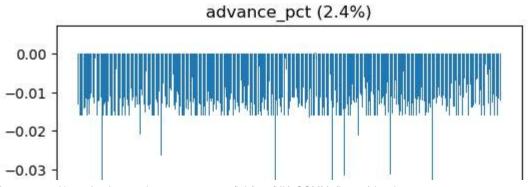


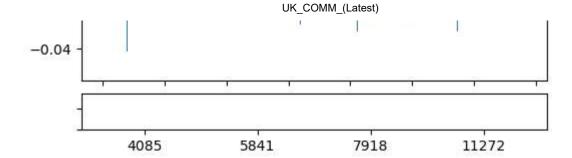


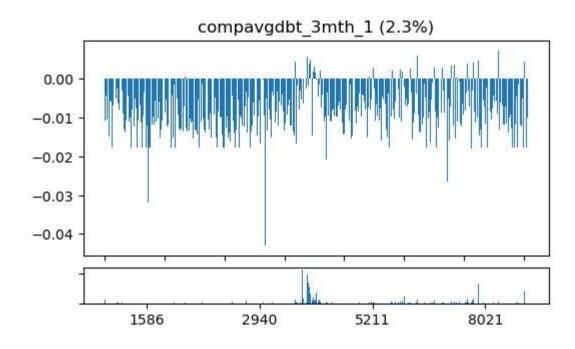


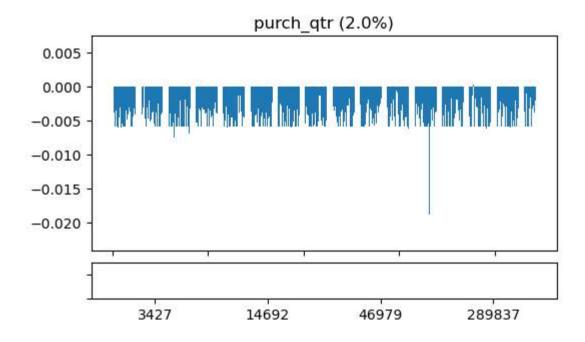


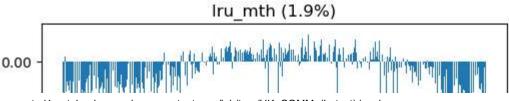


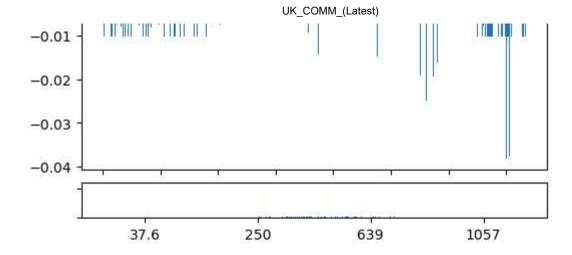


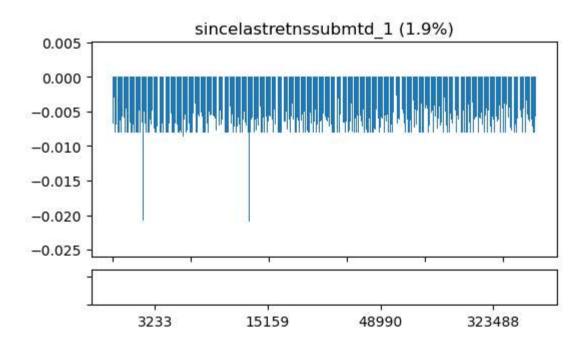


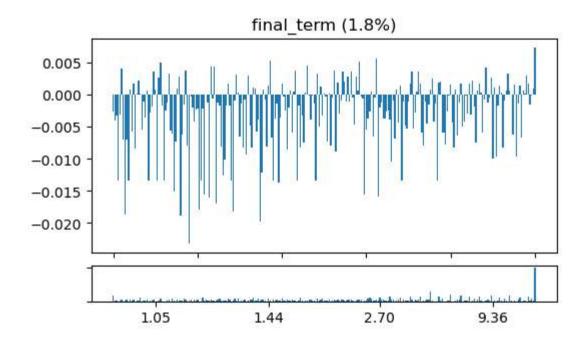


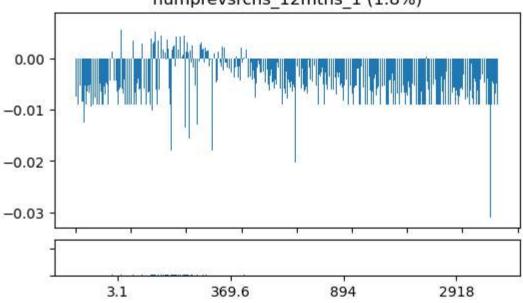


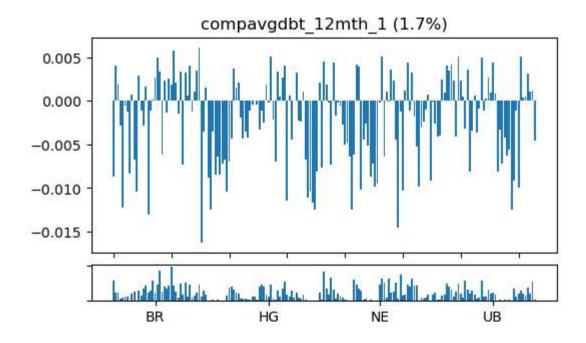


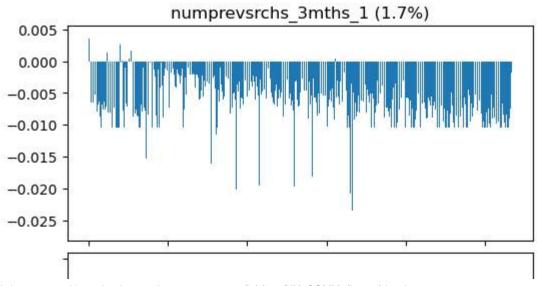


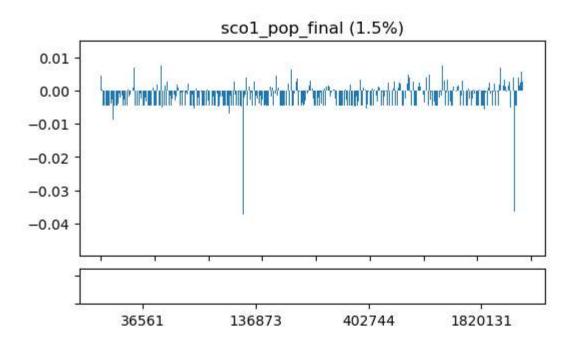


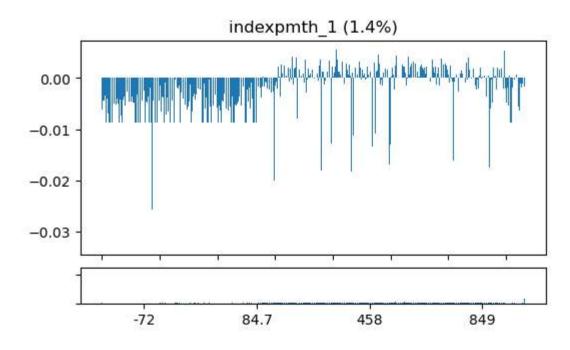


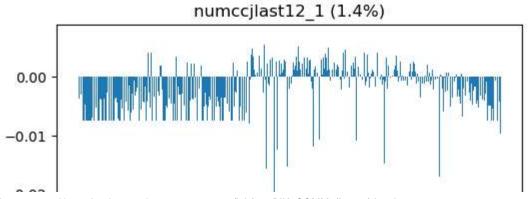


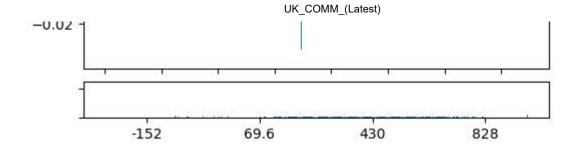


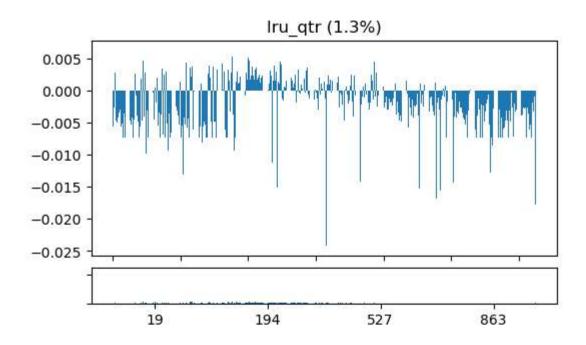


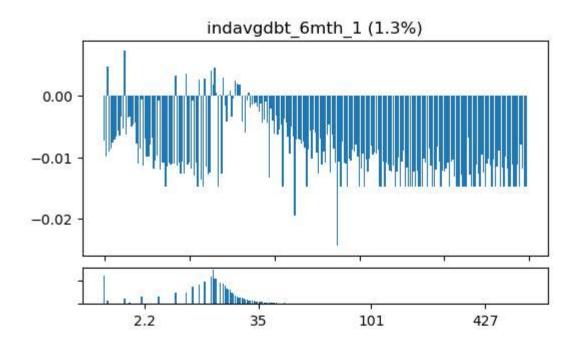


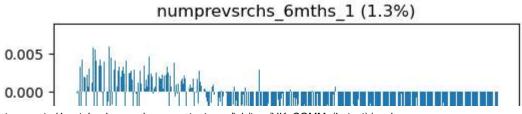


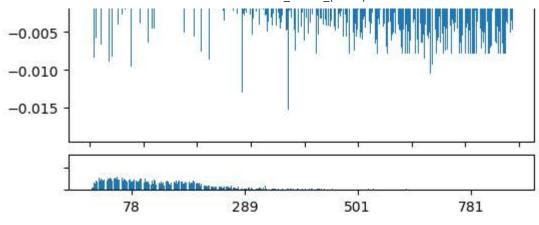


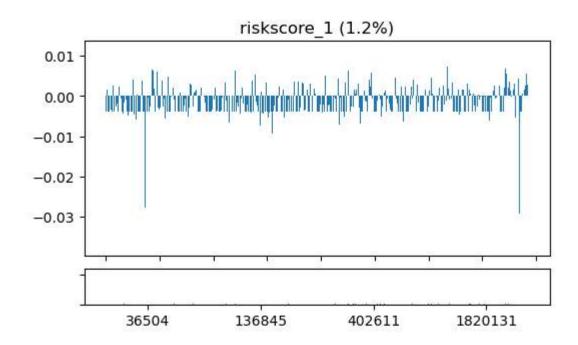


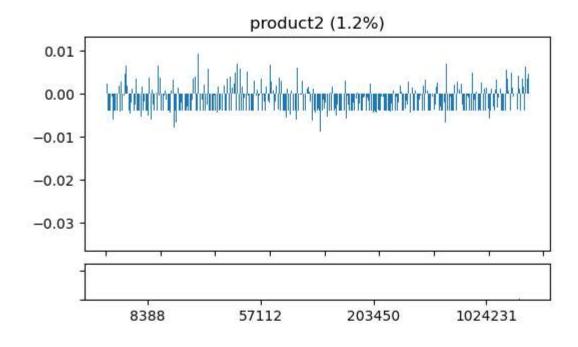


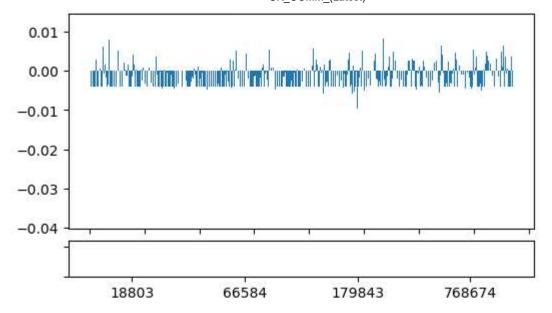


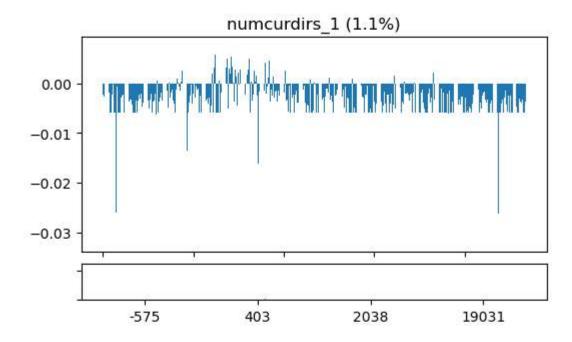


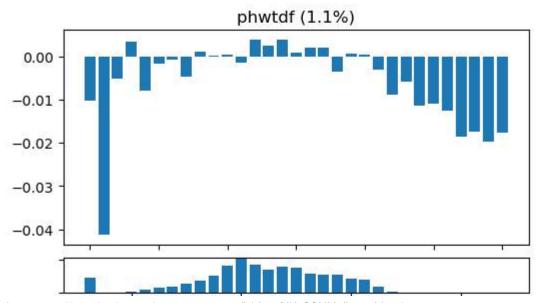


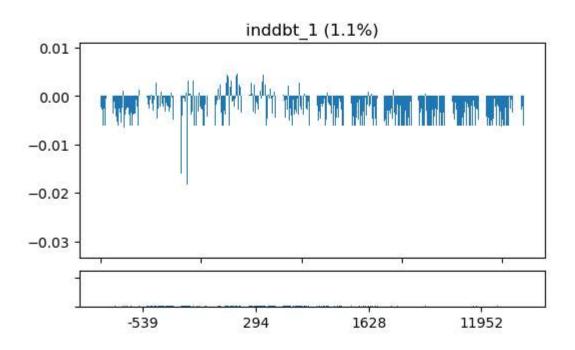


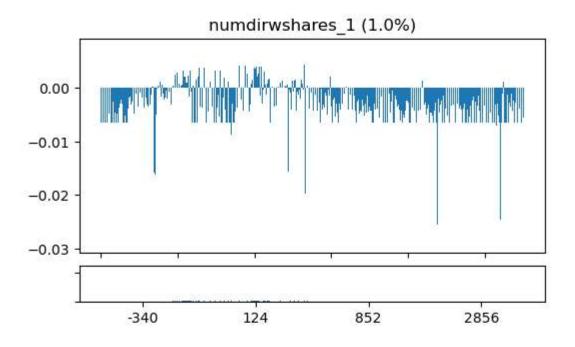


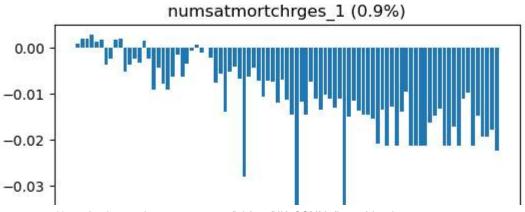


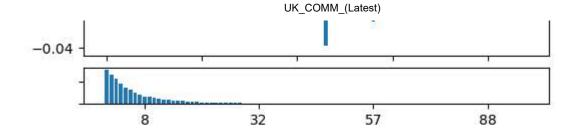


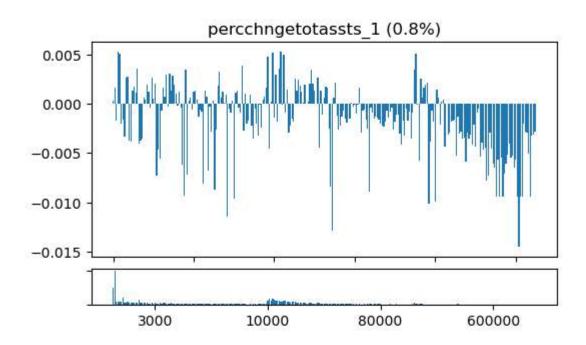


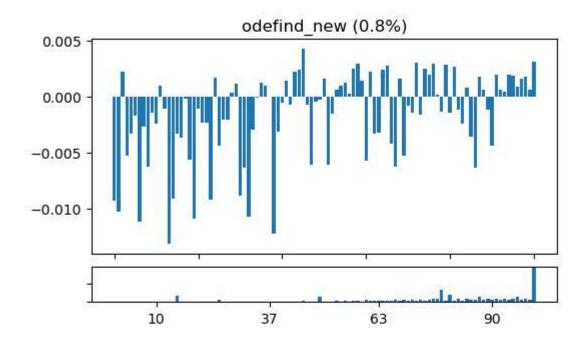




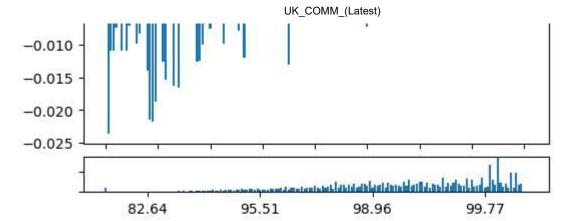


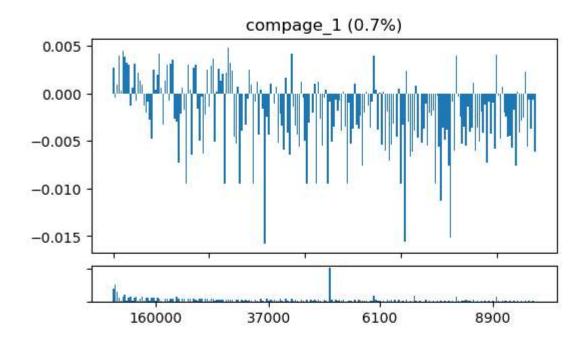


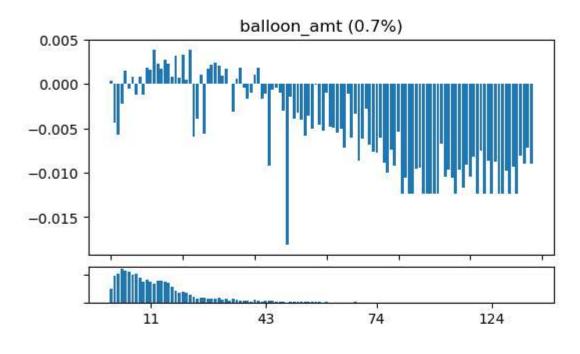






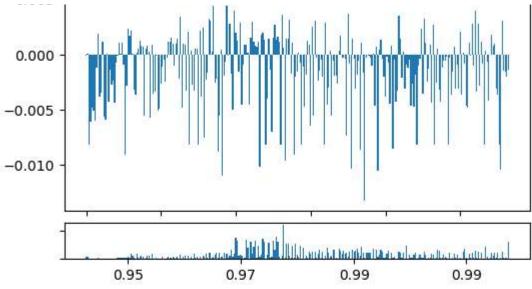


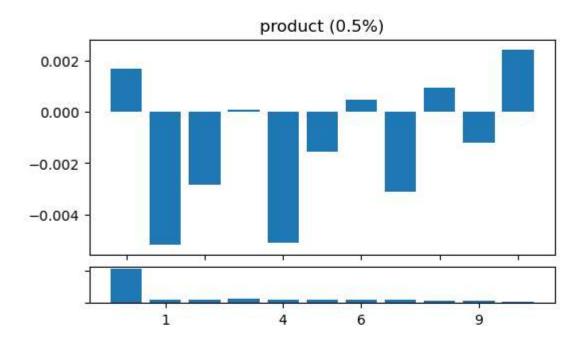


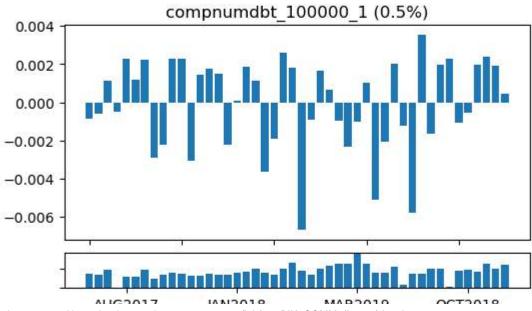


grossassets_1 (0.5%)

0.005 -





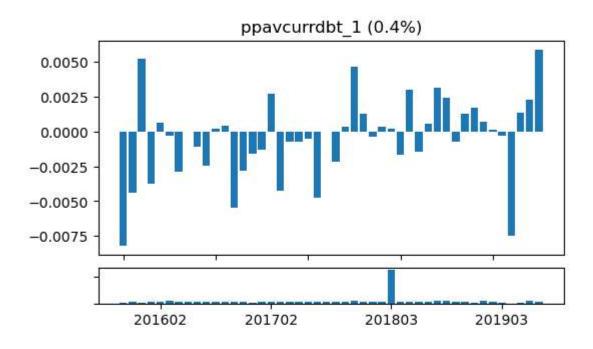


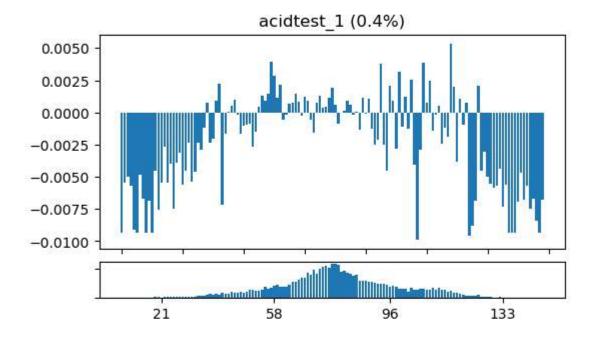
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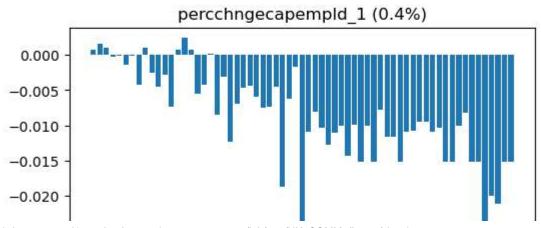
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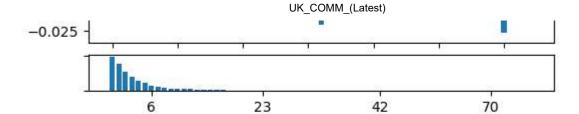
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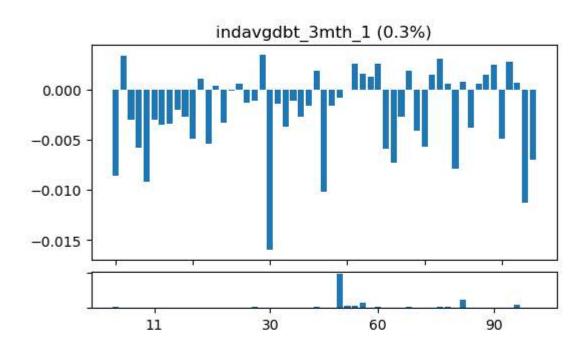
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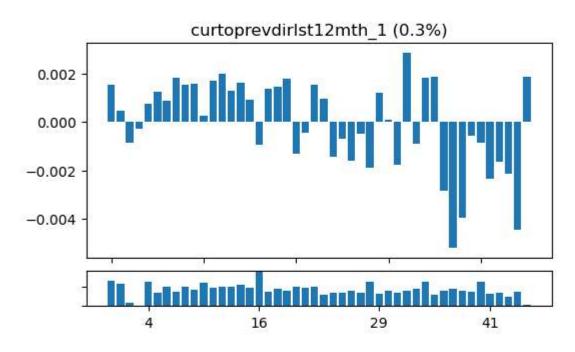


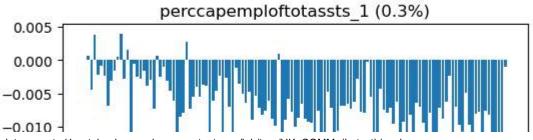


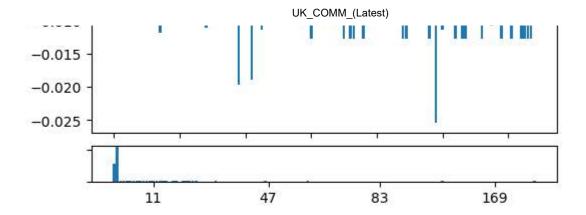


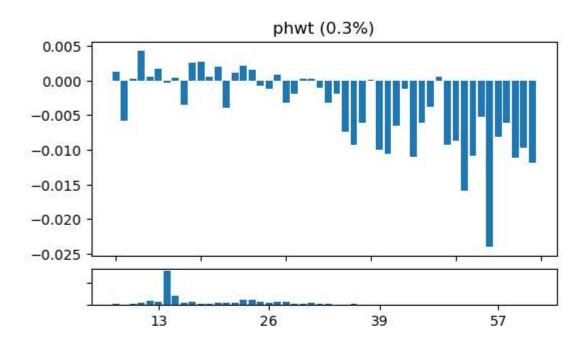


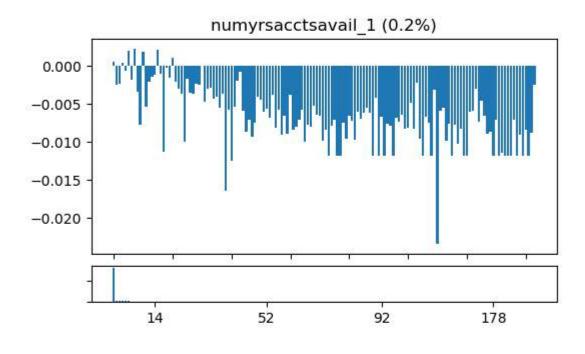




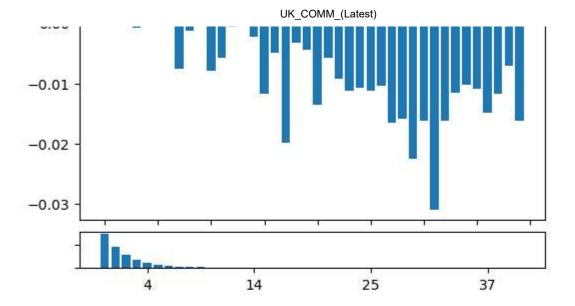


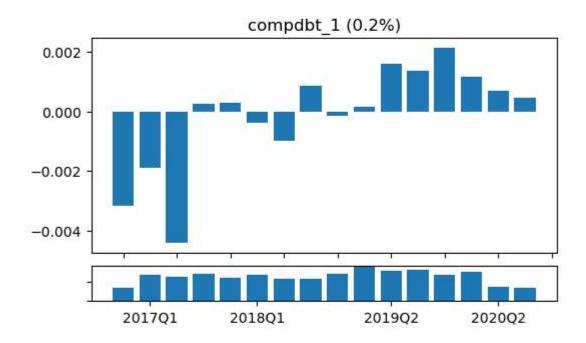


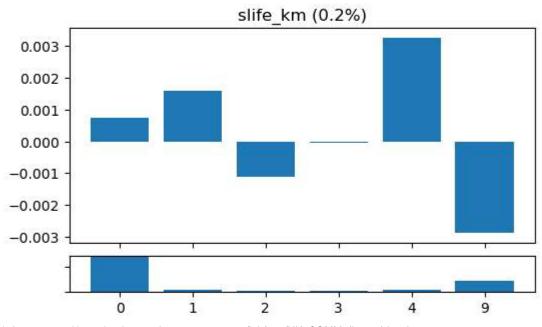


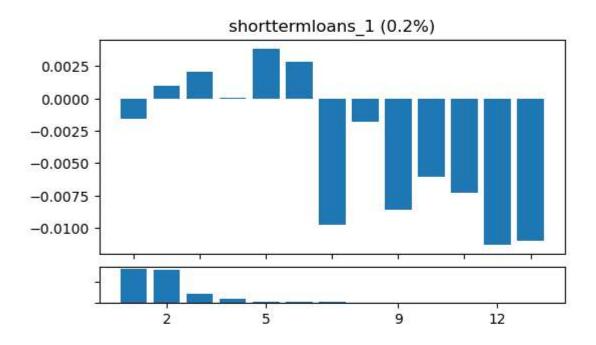


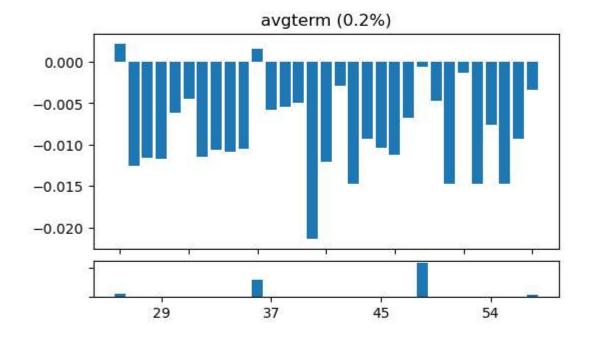


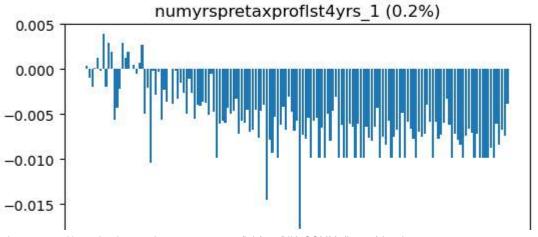


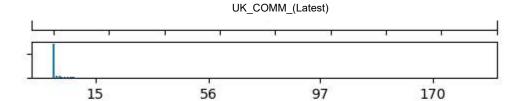


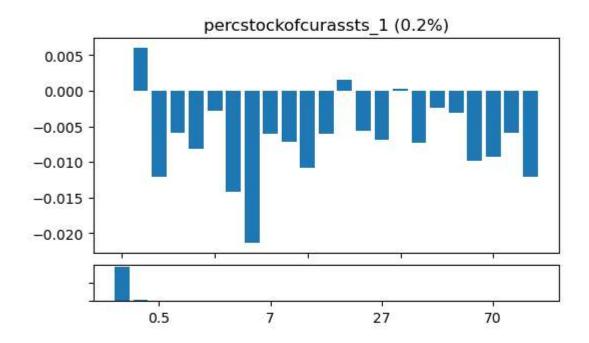


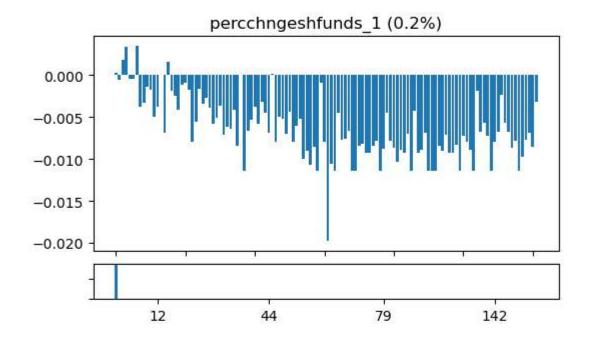


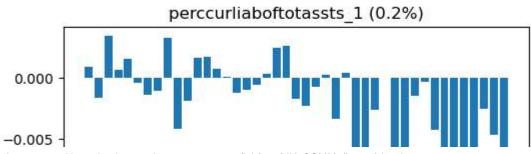


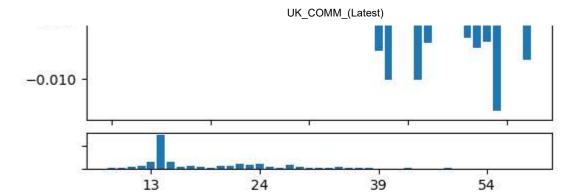


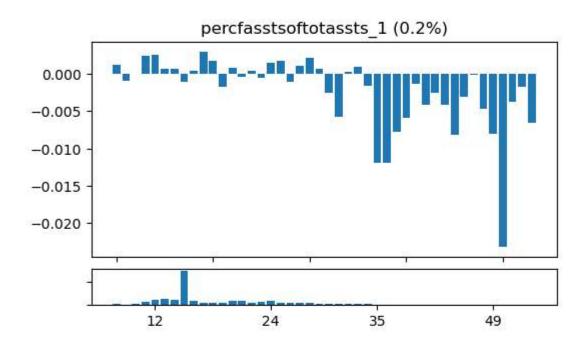


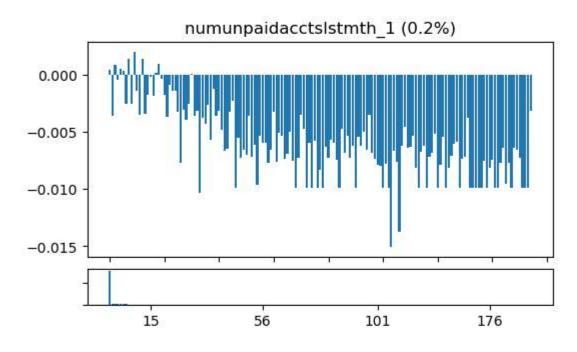




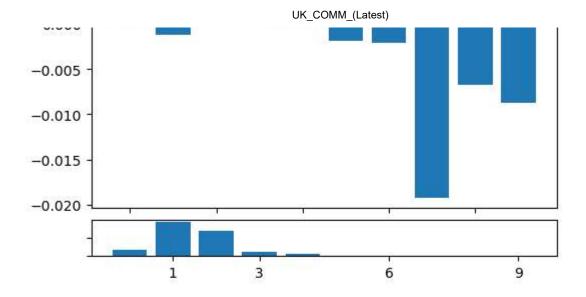


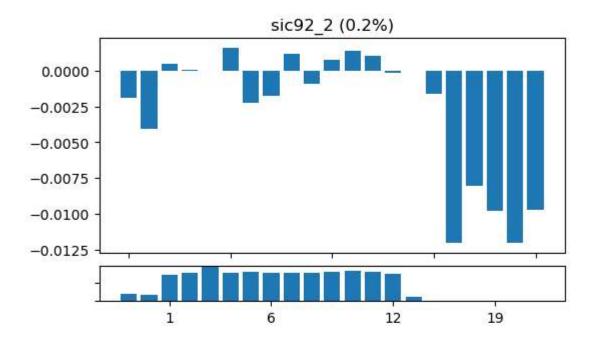


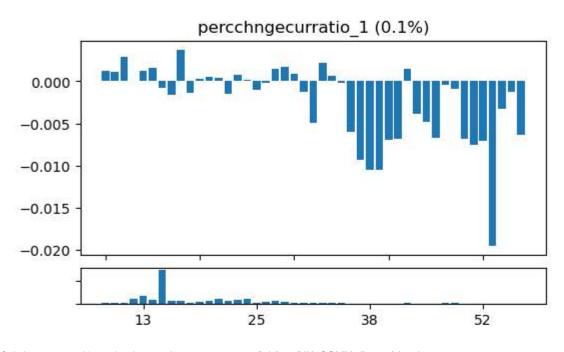


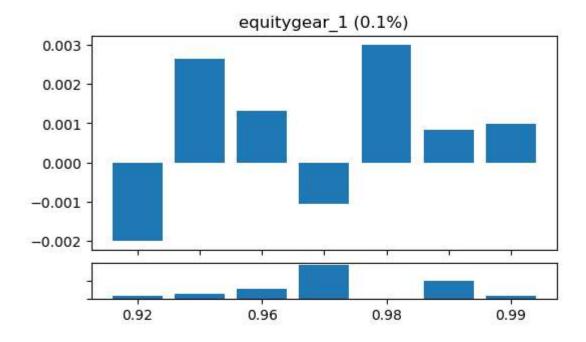


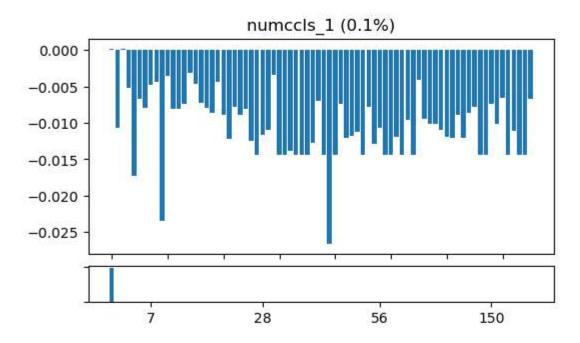


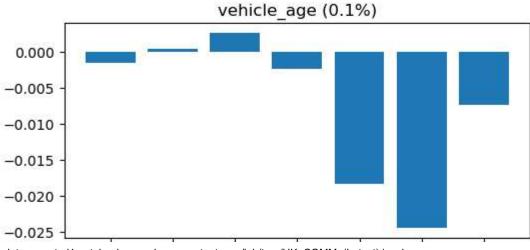


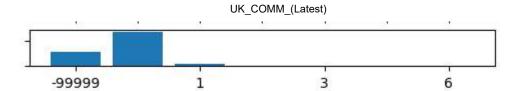


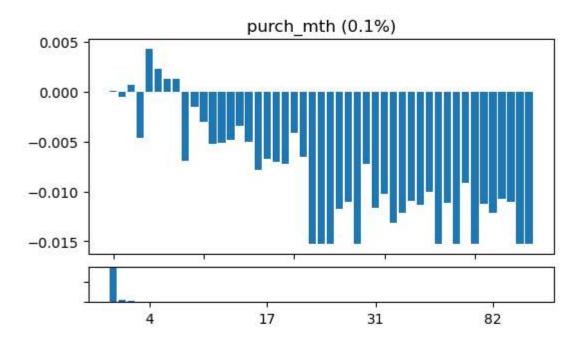


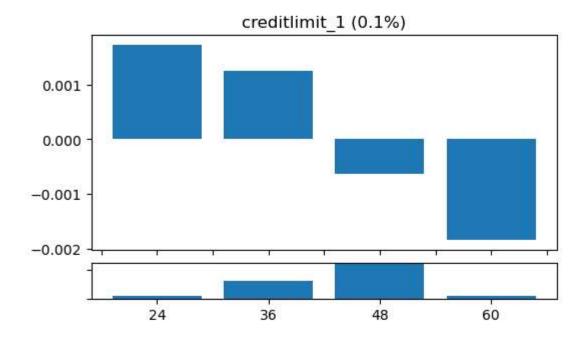


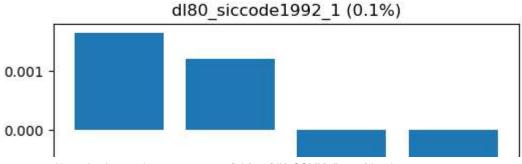


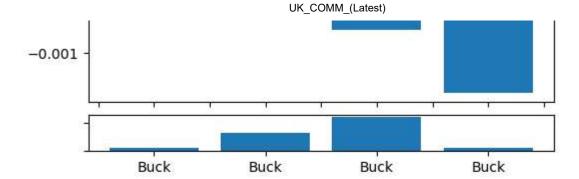


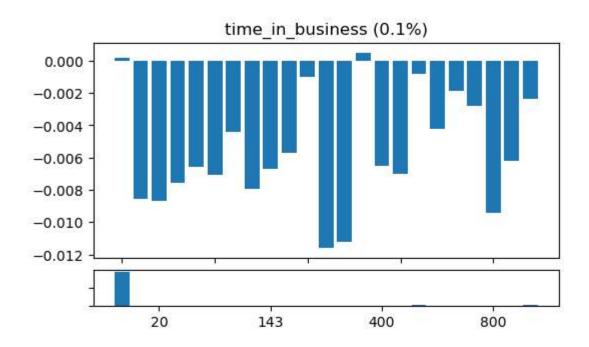


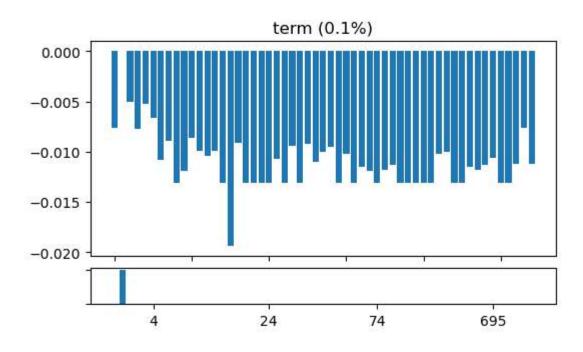


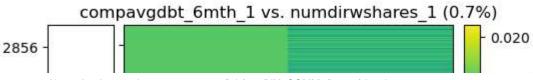


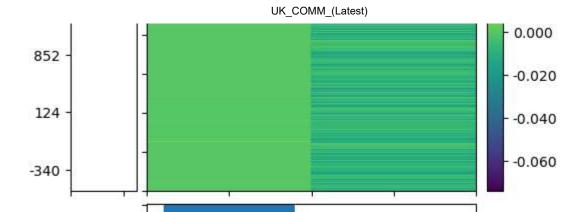




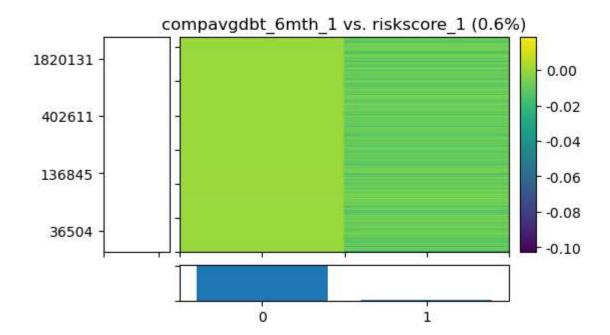




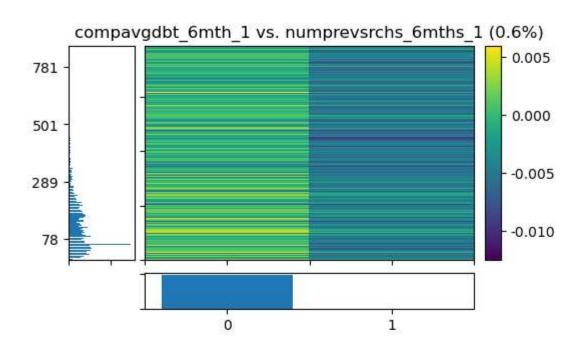


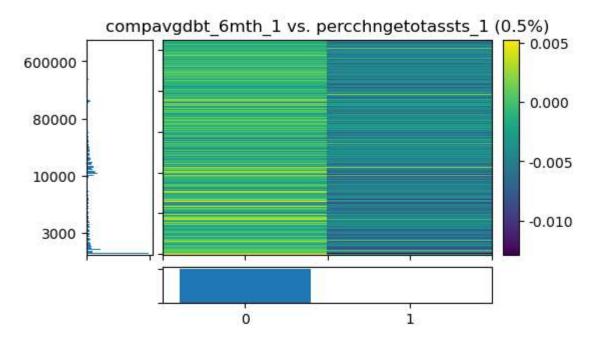


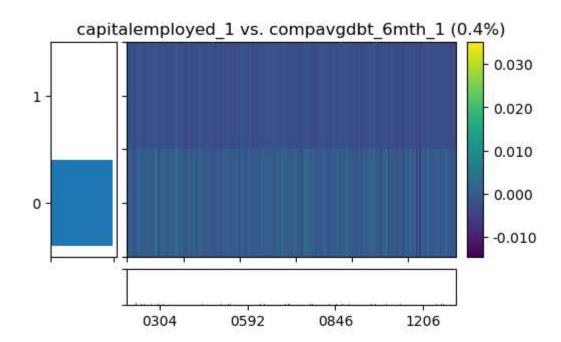
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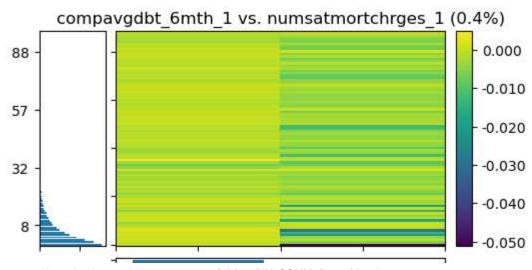


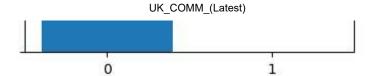
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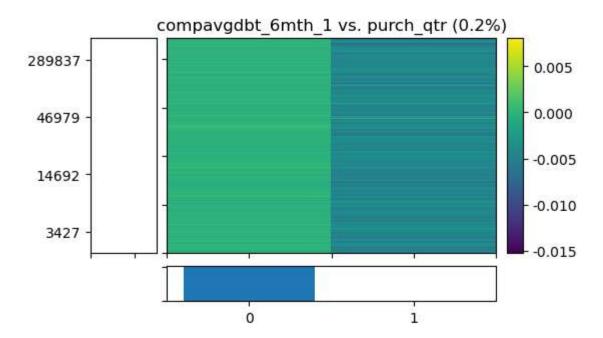


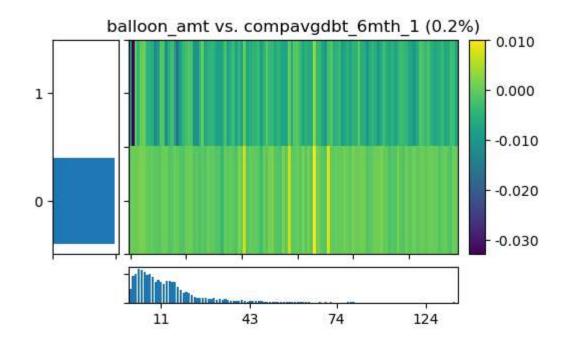


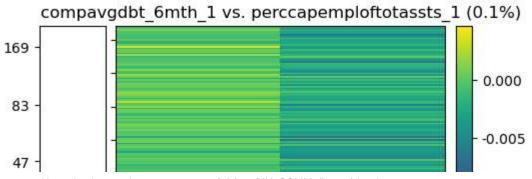


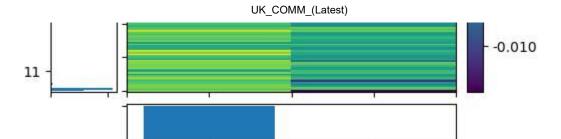






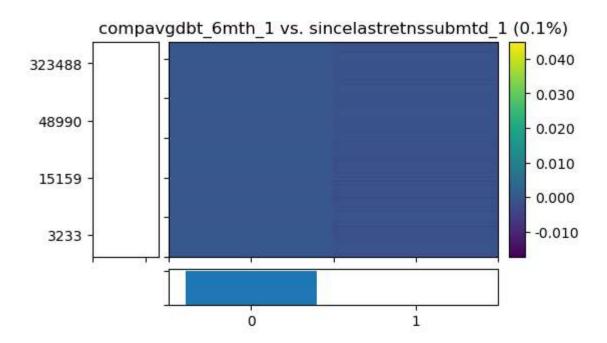


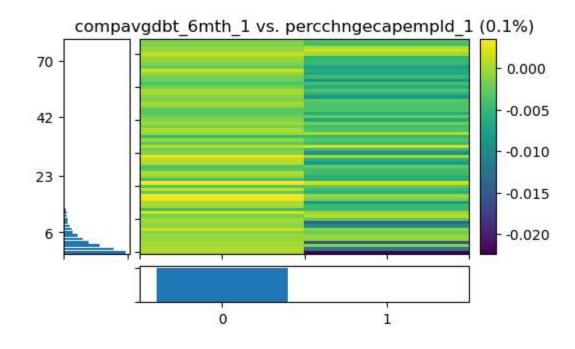




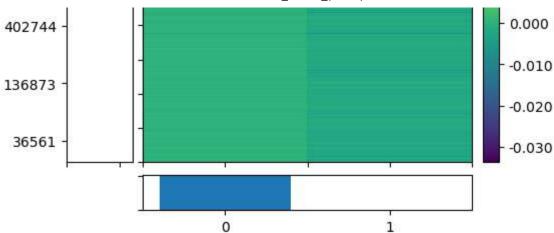
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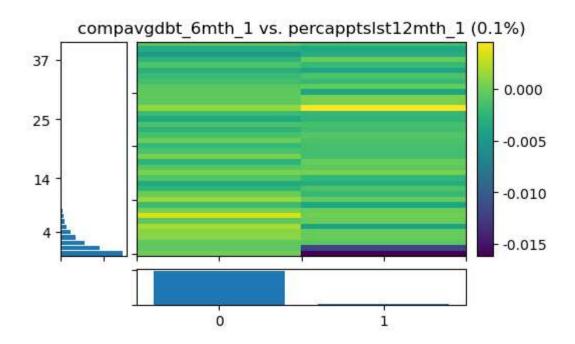
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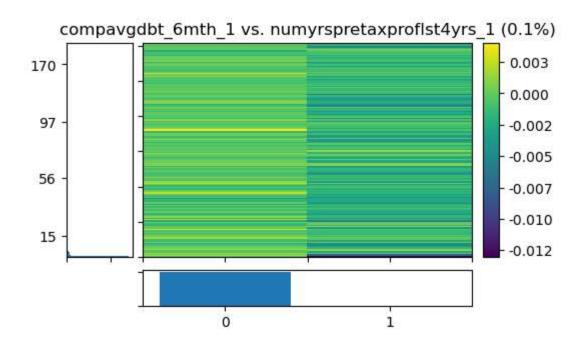


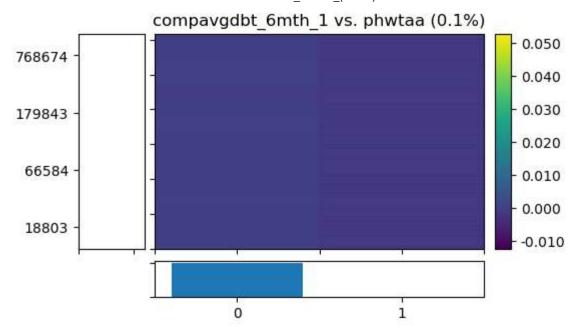


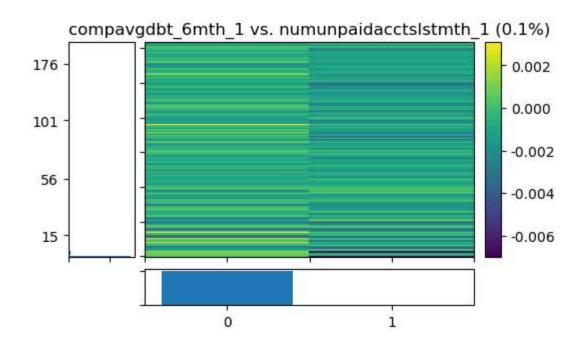


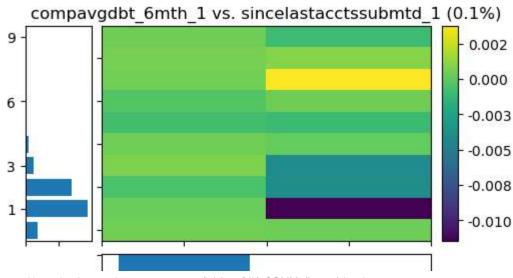


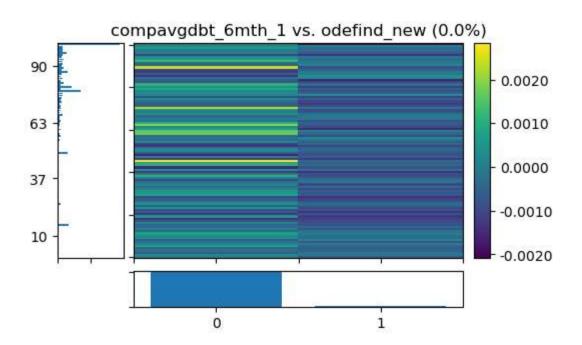


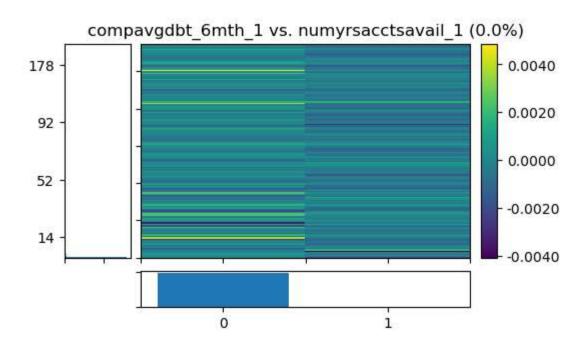


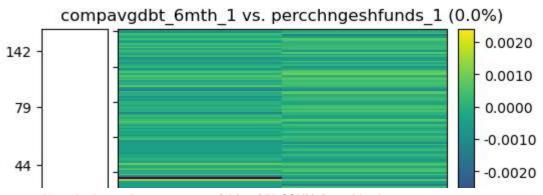


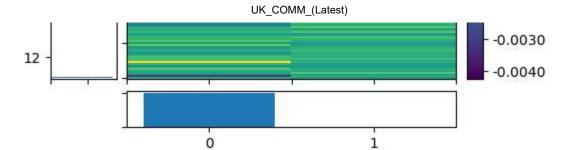


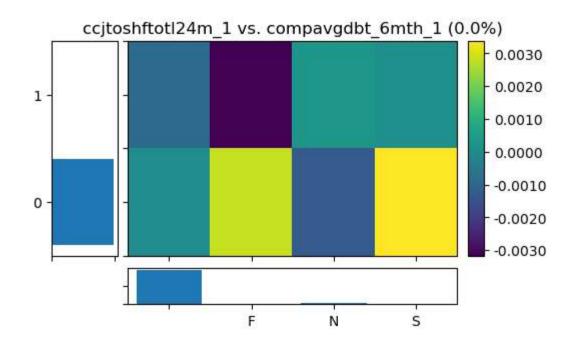


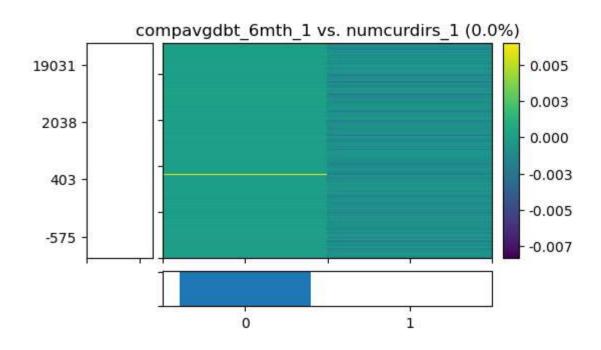




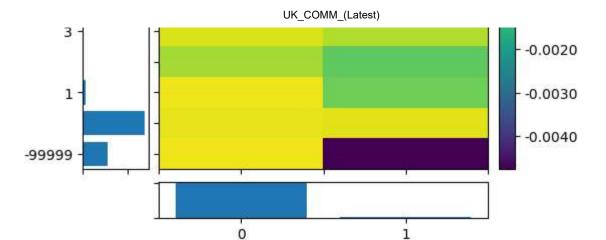


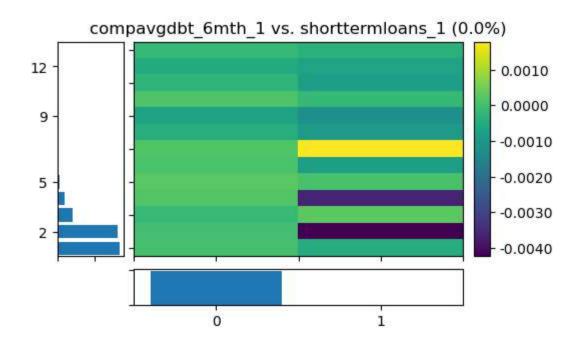


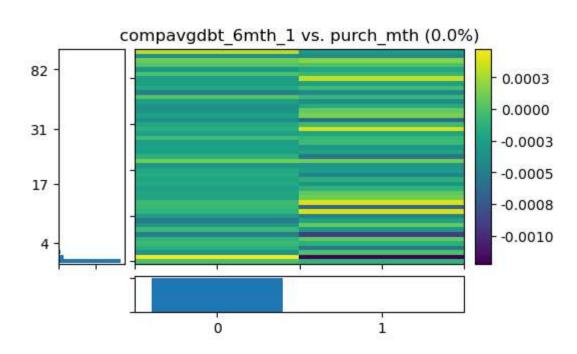


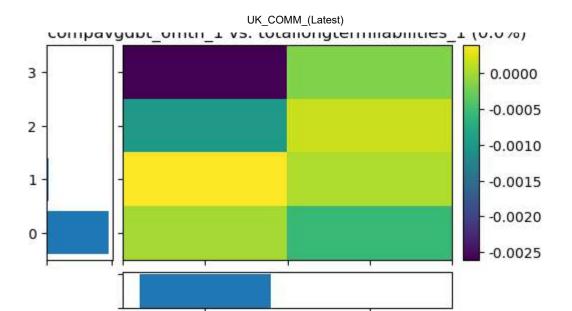






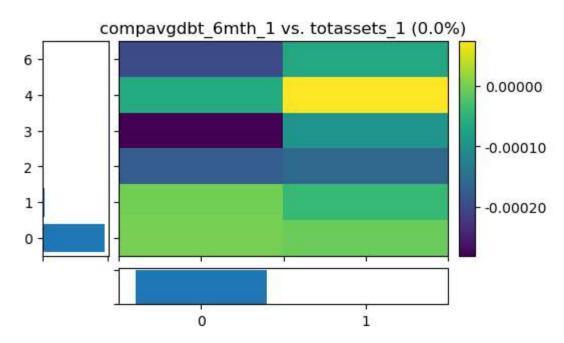


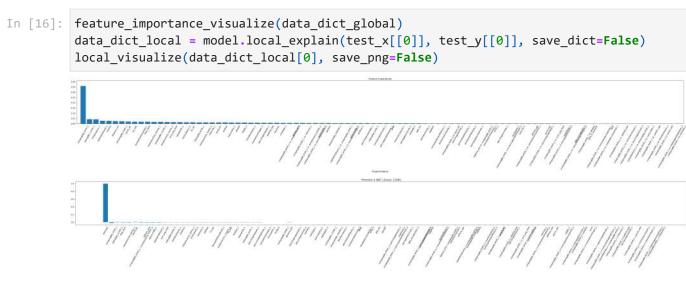




1

0





```
In [ ]: pred_train = model.predict(train_x)
    pred_test = model.predict(test_x)
    print(r2_score(test_y, pred_test))
    print(r2_score(train_y, pred_train))
In [ ]:
```