

reg_year_r_random_points_new_resampled

March 3, 2024

0.1 Importing

```
[ ]: import xarray as xr
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt

from sklearn.model_selection import train_test_split
from sklearn import preprocessing

from sklearn.neural_network import MLPRegressor
from sklearn.ensemble import BaggingRegressor

from sklearn.metrics import mean_squared_error as mse

from tqdm.auto import tqdm

import dill
import random

import salishsea_tools.viz_tools as sa_vi
```

0.2 Datasets Preparation

```
[ ]: def datasets_preparation(dataset):

    drivers = np.stack([np.ravel(dataset['Temperature_(0m-15m)']),
                        np.ravel(dataset['Temperature_(15m-100m)']), np.
↪ ravel(dataset['Salinity_(0m-15m)']),
                        np.ravel(dataset['Salinity_(15m-100m)'])])
    indx = np.where(~np.isnan(drivers).any(axis=0))
    drivers = drivers[:,indx[0]]

    diat = np.ravel(dataset['Diatom'])
    diat = diat[indx[0]]

    return(drivers, diat, indx)
```

0.3 Regressor

```
[ ]: def regressor (inputs, targets):

    inputs = inputs.transpose()

    # Regressor
    scale = preprocessing.StandardScaler()
    inputs = scale.fit_transform(inputs)
    X_train, _, y_train, _ = train_test_split(inputs, targets, train_size=0.35)

    drivers = None
    diat = None

    inputs = None
    targets = None

    model = MLPRegressor(hidden_layer_sizes=200, alpha=0.002)
    regr = BaggingRegressor(model, n_estimators=12, n_jobs=4).fit(X_train,
↪y_train)

    return (regr)
```

0.4 Regressor 2

```
[ ]: def regressor2 (inputs, targets, variable_name):

    inputs = inputs.transpose()

    # Regressor
    scale = preprocessing.StandardScaler()
    inputs2 = scale.fit_transform(inputs)

    outputs_test = regr.predict(inputs2)

    m = scatter_plot(targets, outputs_test, variable_name)
    r = np.round(np.corrcoef(targets, outputs_test)[0][1],3)
    rms = mse(targets, outputs_test)

    return (r, rms, m)
```

0.5 Regressor 3

```
[ ]: def regressor3 (inputs, targets):

    inputs = inputs.transpose()
```

```

# Regressor
scale = preprocessing.StandardScaler()
inputs2 = scale.fit_transform(inputs)

outputs_test = regr.predict(inputs2)

# compute slope m and intercept b
m, b = np.polyfit(targets, outputs_test, deg=1)

r = np.round(np.corrcoef(targets, outputs_test)[0][1],3)
rms = mse(targets, outputs_test)

return (r, rms, m)

```

0.6 Regressor 4

```

[ ]: def regressor4 (inputs, targets, variable_name):

    inputs = inputs.transpose()

    # Regressor
    scale = preprocessing.StandardScaler()
    inputs2 = scale.fit_transform(inputs)

    outputs = regr.predict(inputs2)

    # Post processing
    indx2 = np.full((len(diat_i.y)*len(diat_i.x)),np.nan)
    indx2[indx[0]] = outputs
    model = np.reshape(indx2,(len(diat_i.y),len(diat_i.x)))

    m = scatter_plot(targets, outputs, variable_name + str(dates[i].date()))

    # Preparation of the dataarray
    model = xr.DataArray(model,
        coords = {'y': diat_i.y, 'x': diat_i.x},
        dims = ['y','x'],
        attrs=dict( long_name = variable_name + "Concentration",
            units="mmol m-2"),)

    plotting3(targets, model, diat_i, variable_name)

```

1 Printing

```
[ ]: def printing (targets, outputs, m):  
  
    print ('The amount of data points is', outputs.size)  
    print ('The slope of the best fitting line is ', np.round(m,3))  
    print ('The correlation coefficient is:', np.round(np.corrcoef(targets,   
→outputs)[0][1],3))  
    print (' The mean square error is:', np.round(mse(targets,outputs),5))
```

1.1 Scatter Plot

```
[ ]: def scatter_plot(targets, outputs, variable_name):  
  
    # compute slope m and intercept b  
    m, b = np.polyfit(targets, outputs, deg=1)  
  
    printing(targets, outputs, m)  
  
    fig, ax = plt.subplots(2, figsize=(5,10), layout='constrained')  
  
    ax[0].scatter(targets,outputs, alpha = 0.2, s = 10)  
  
    lims = [np.min([ax[0].get_xlim(), ax[0].get_ylim()]),  
            np.max([ax[0].get_xlim(), ax[0].get_ylim()])]  
  
    # plot fitted y = m*x + b  
    ax[0].axline(xy1=(0, b), slope=m, color='r')  
  
    ax[0].set_xlabel('targets')  
    ax[0].set_ylabel('outputs')  
    ax[0].set_xlim(lims)  
    ax[0].set_ylim(lims)  
    ax[0].set_aspect('equal')  
  
    ax[0].plot(lims, lims,linestyle = '--',color = 'k')  
  
    h = ax[1].hist2d(targets,outputs, bins=100, cmap='jet',  
                    range=[lims,lims], cmin=0.1, norm='log')  
  
    ax[1].plot(lims, lims,linestyle = '--',color = 'k')  
  
    # plot fitted y = m*x + b  
    ax[1].axline(xy1=(0, b), slope=m, color='r')  
  
    ax[1].set_xlabel('targets')  
    ax[1].set_ylabel('outputs')
```

```

ax[1].set_aspect('equal')

fig.colorbar(h[3],ax=ax[1], location='bottom')

fig.suptitle(variable_name)

plt.show()

return (m)

```

1.2 Plotting

```

[ ]: def plotting(variable, name):

    plt.plot(years,variable, marker = '.', linestyle = '')
    plt.legend(['diatom','flagellate'])
    plt.xlabel('Years')
    plt.ylabel(name)
    plt.show()

```

1.3 Plotting 2

```

[ ]: def plotting2(variable,title):

    fig, ax = plt.subplots()

    scatter= ax.scatter(dates,variable, marker='.', c=pd.DatetimeIndex(dates).
↳month)

    ax.legend(handles=scatter.legend_elements()[0],
↳labels=['February','March','April'])
    fig.suptitle('Daily ' + title + ' (15 Feb - 30 Apr)')

    fig.show()

```

1.4 Plotting 3

```

[ ]: def plotting3(targets, model, variable, variable_name):

    fig, ax = plt.subplots(2,2, figsize = (10,15))

    cmap = plt.get_cmap('cubehelix')
    cmap.set_bad('gray')

    variable.plot(ax=ax[0,0], cmap=cmap, vmin = targets.min(), vmax =targets.
↳max(), cbar_kwargs={'label': variable_name + ' Concentration [mmol m-2]'})

```

```

    model.plot(ax=ax[0,1], cmap=cmap, vmin = targets.min(), vmax = targets.
↪max(), cbar_kwargs={'label': variable_name + ' Concentration [mmol m-2]'})
    ((variable-model) / variable * 100).plot(ax=ax[1,0], cmap=cmap,
↪cbar_kwargs={'label': variable_name + ' Concentration [percentage]'})

plt.subplots_adjust(left=0.1,
    bottom=0.1,
    right=0.95,
    top=0.95,
    wspace=0.35,
    hspace=0.35)

sa_vi.set_aspect(ax[0,0])
sa_vi.set_aspect(ax[0,1])
sa_vi.set_aspect(ax[1,0])

ax[0,0].title.set_text(variable_name + ' (targets)')
ax[0,1].title.set_text(variable_name + ' (outputs)')
ax[1,0].title.set_text('targets - outputs')
ax[1,1].axis('off')

fig.suptitle(str(dates[i].date()))

plt.show()

```

1.5 Training (Random Points)

```

[ ]: ds = xr.open_dataset('/data/ibougoudis/MOAD/files/integrated_model_var_old.nc')

ds = ds.isel(time_counter = (np.arange(0, len(ds.Diatom.time_counter),2)),
    y=(np.arange(ds.y[0], ds.y[-1], 5)),
    x=(np.arange(ds.x[0], ds.x[-1], 5)))

dates = pd.DatetimeIndex(ds['time_counter'].values)

drivers, diat, _ = datasets_preparation(ds)

regr = regressor(drivers, diat)

```

1.6 Other Years (Anually)

```
[ ]: years = range (2007,2024)

r_all = []
rms_all = []
slope_all = []

for year in tqdm(range (2007,2024)):

    dataset = ds.sel(time_counter=str(year))

    drivers, diat, _ = datasets_preparation(dataset)

    r, rms, m = regressor2(drivers, diat, 'Diatom ' + str(year))

    r_all.append(r)
    rms_all.append(rms)
    slope_all.append(m)

plotting(np.transpose(r_all), 'Correlation Coefficient')
plotting(np.transpose(rms_all), 'Mean Square Error')
plotting (np.transpose(slope_all), 'Slope of the best fitting line')
```

0%| | 0/17 [00:00<?, ?it/s]

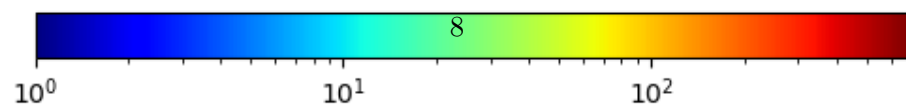
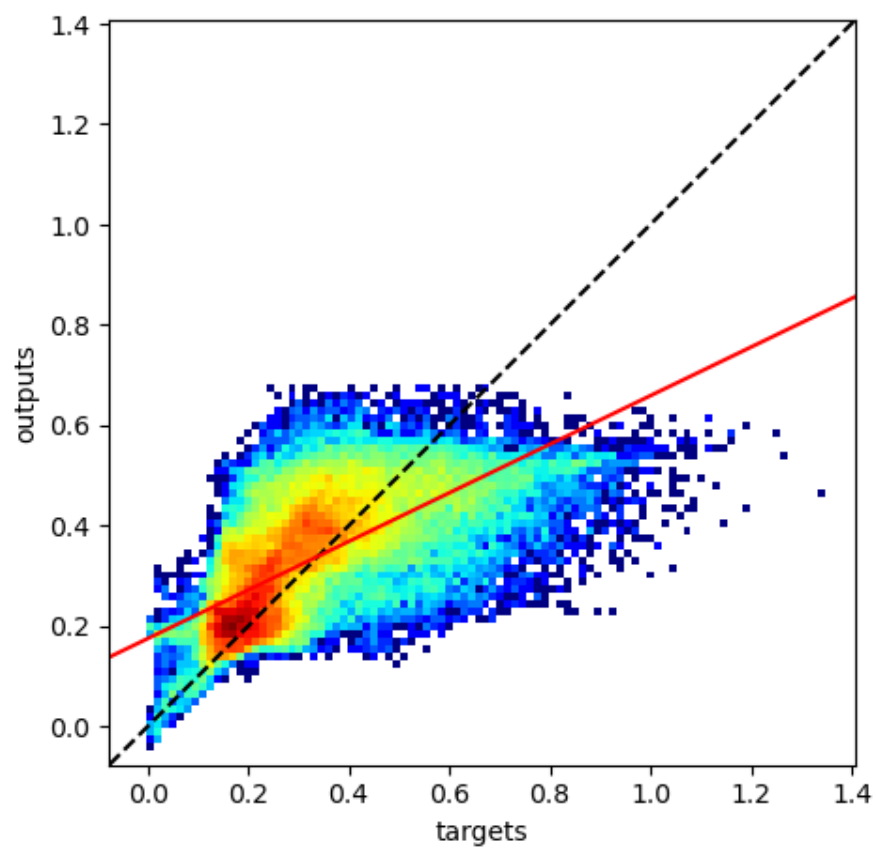
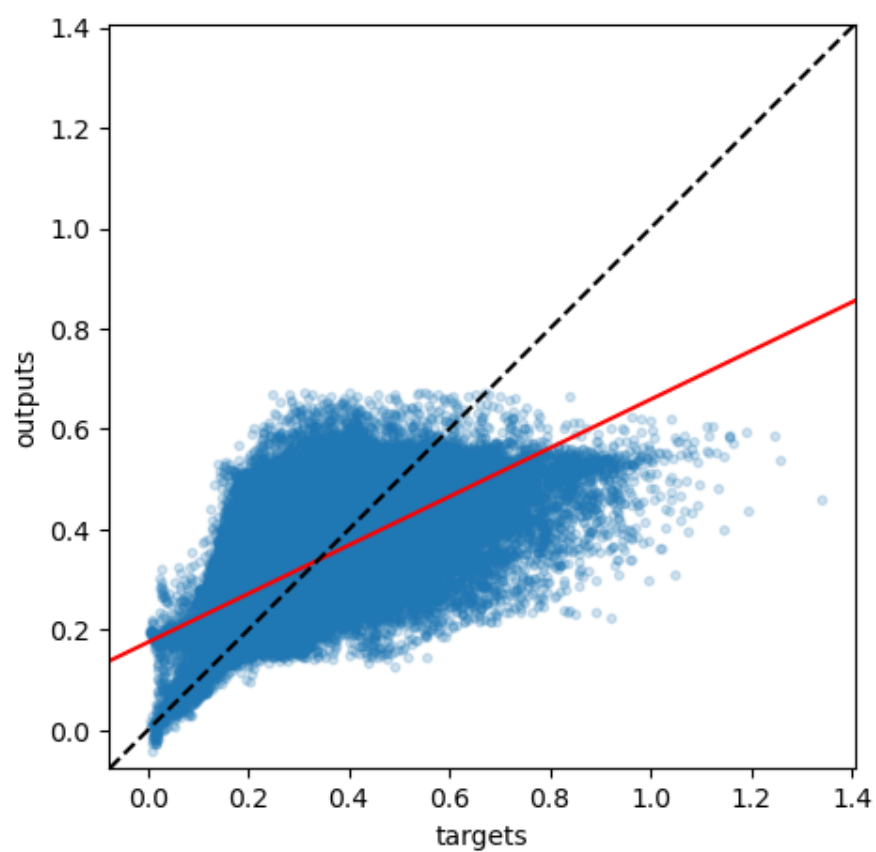
The amount of data points is 70794

The slope of the best fitting line is 0.484

The correlation coefficient is: 0.67

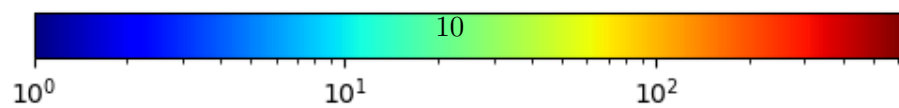
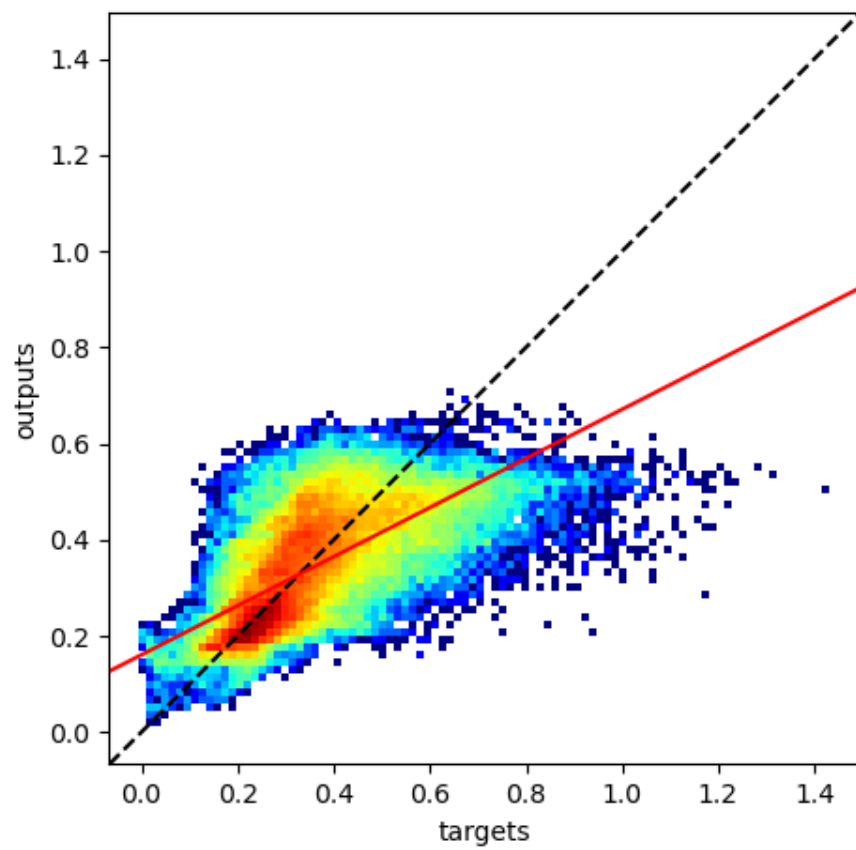
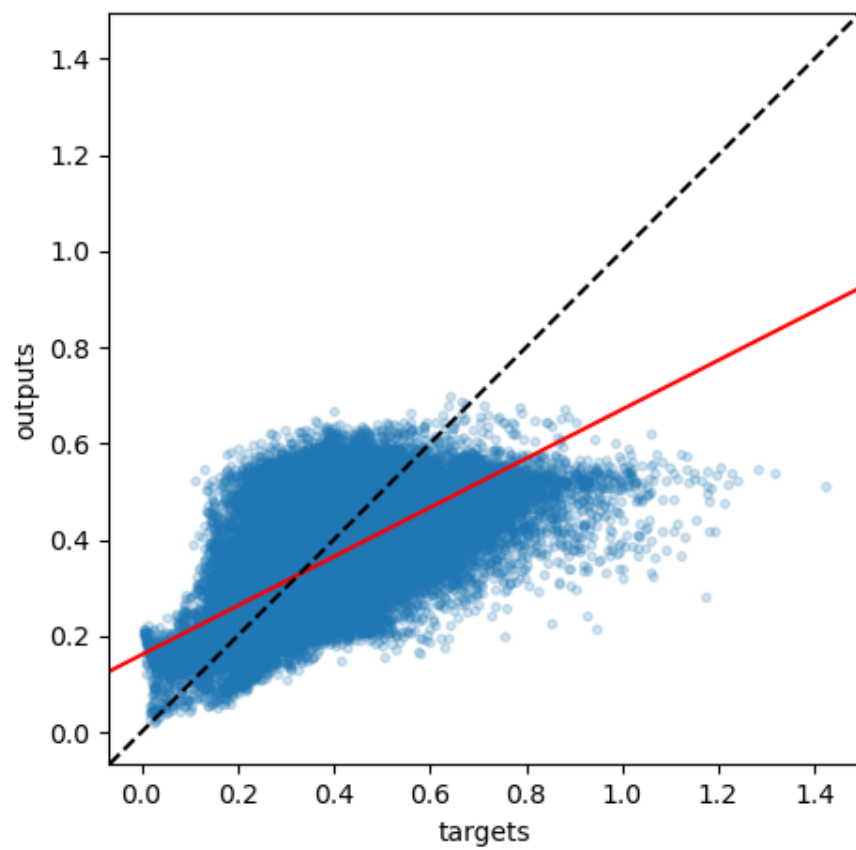
The mean square error is: 0.01434

Diatom 2007



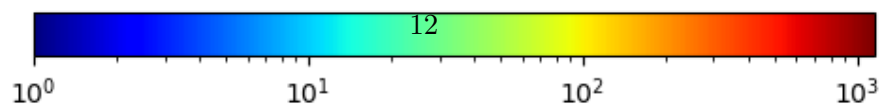
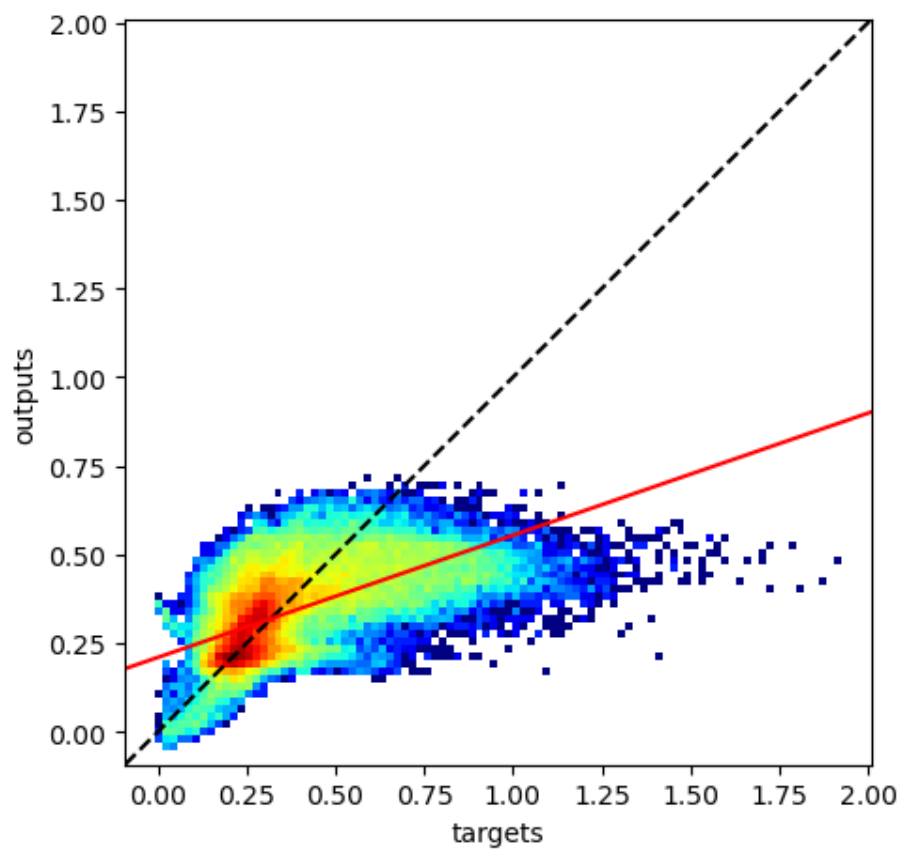
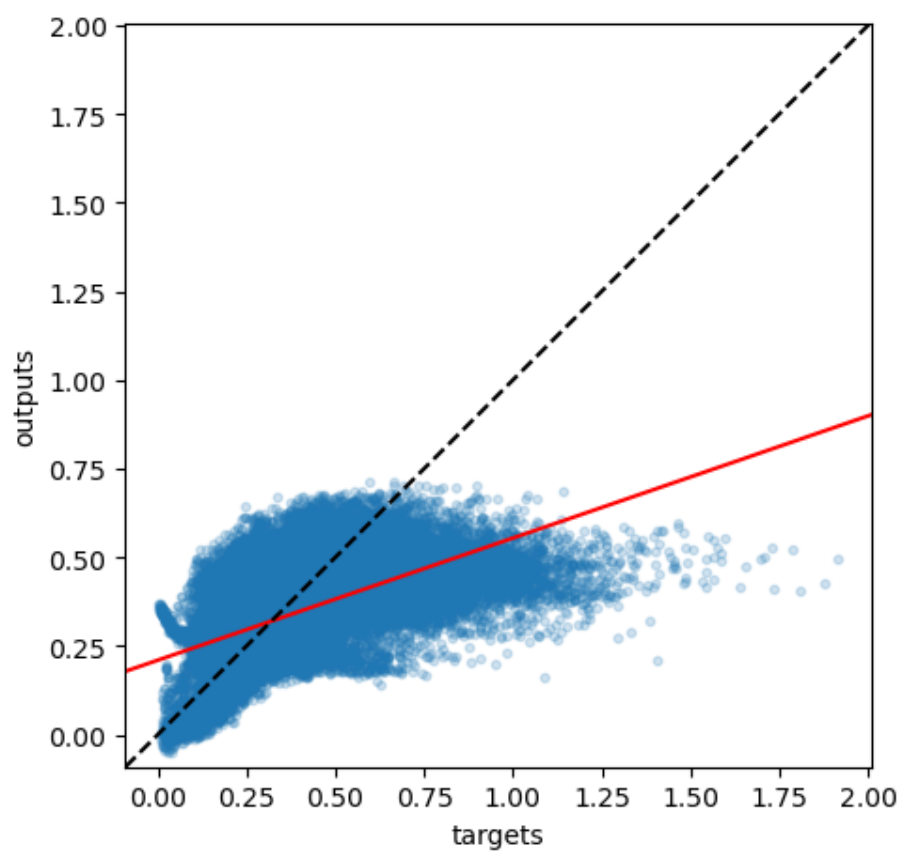
The amount of data points is 70794
The slope of the best fitting line is 0.51
The correlation coefficient is: 0.659
The mean square error is: 0.01228

Diatom 2008



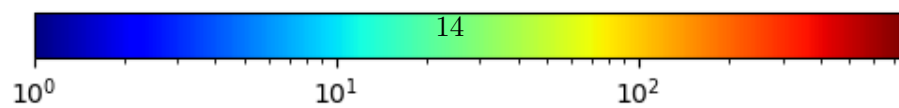
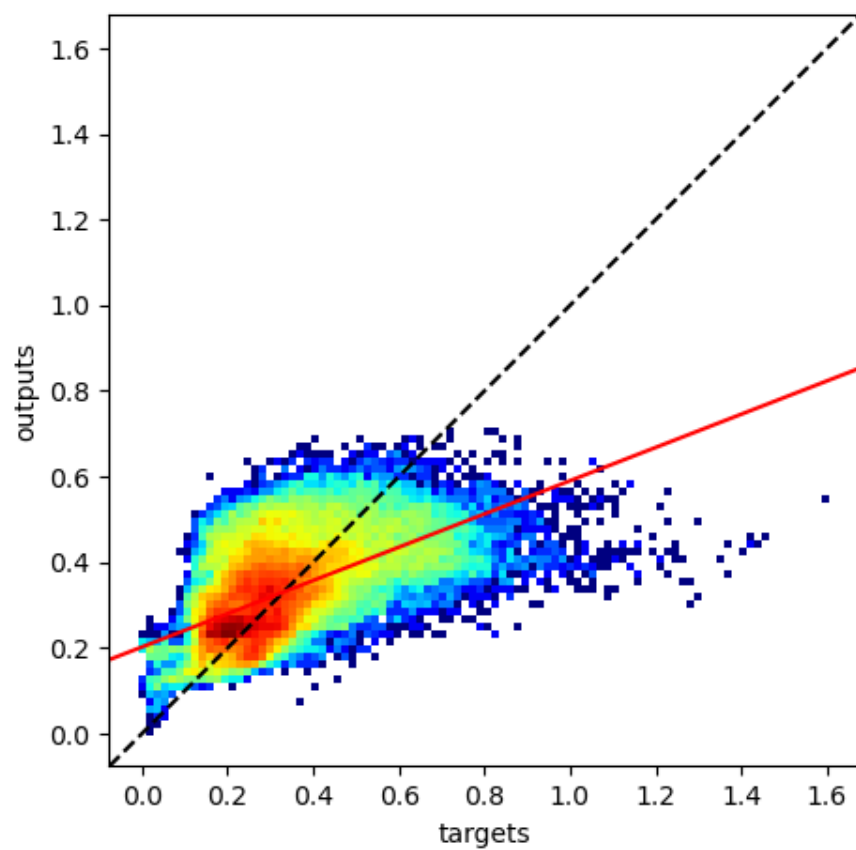
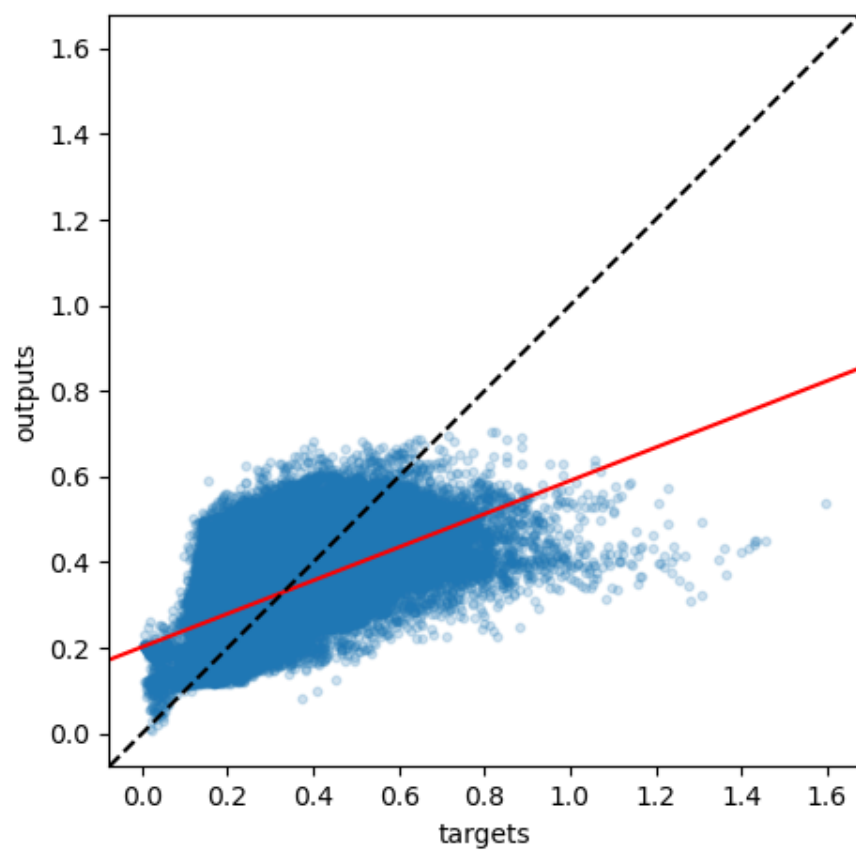
The amount of data points is 68931
The slope of the best fitting line is 0.344
The correlation coefficient is: 0.596
The mean square error is: 0.02519

Diatom 2009



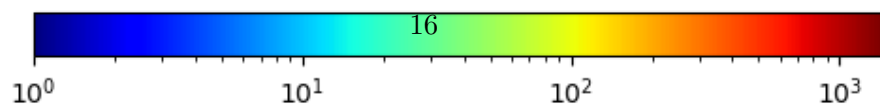
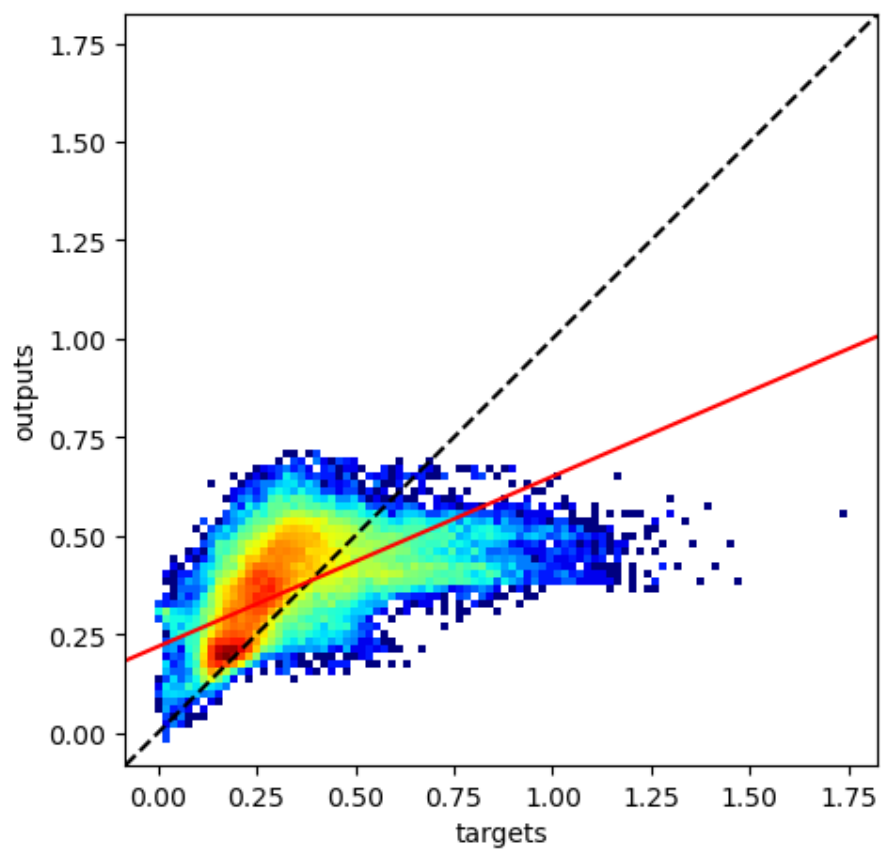
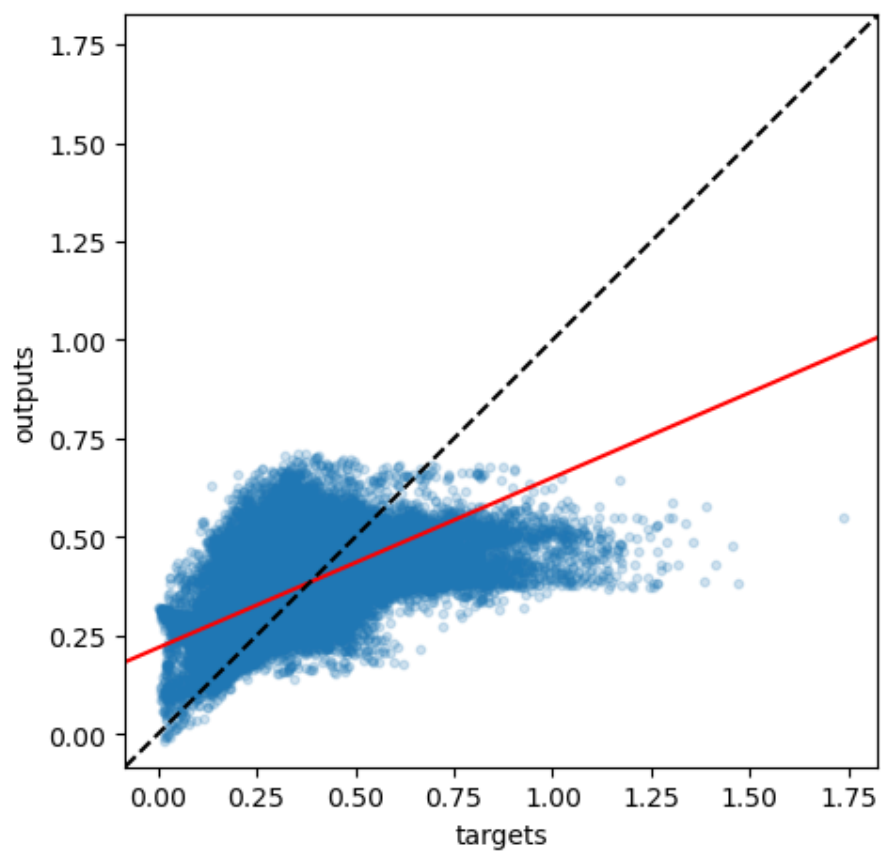
The amount of data points is 70794
The slope of the best fitting line is 0.388
The correlation coefficient is: 0.566
The mean square error is: 0.01468

Diatom 2010



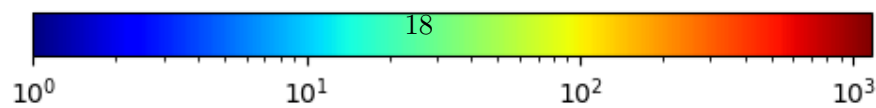
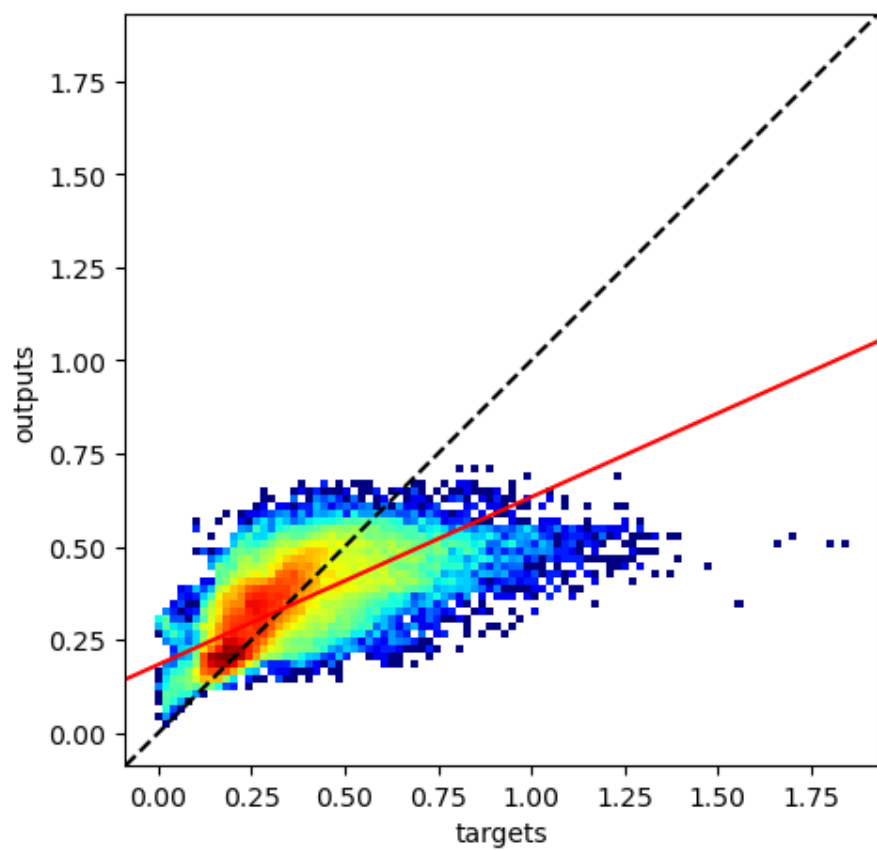
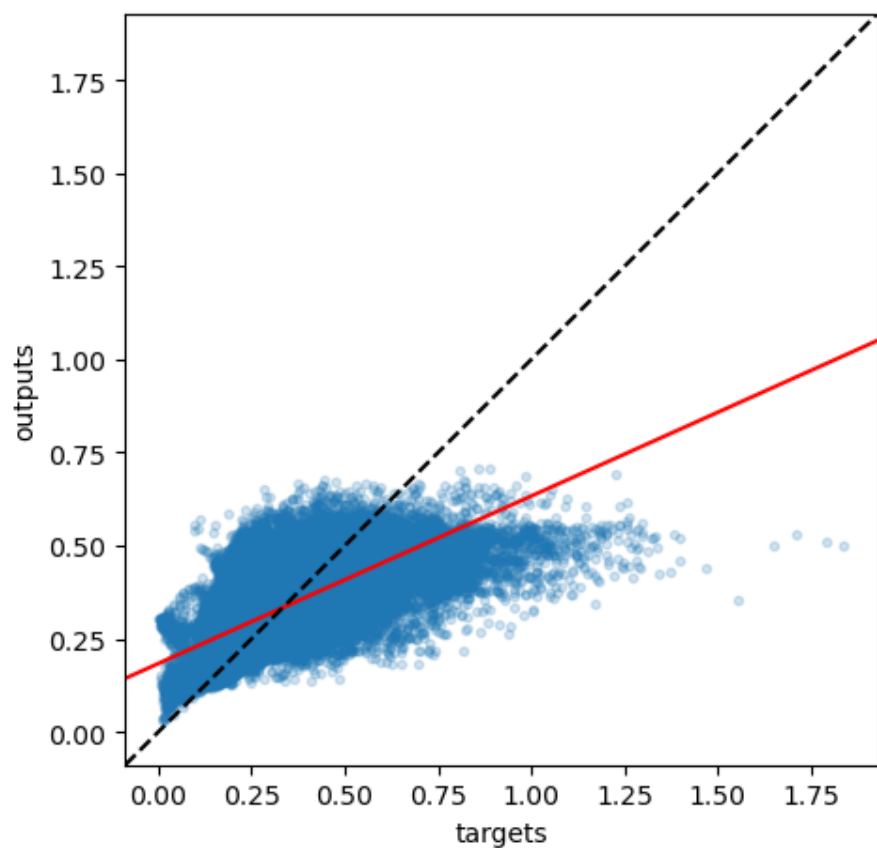
The amount of data points is 68931
The slope of the best fitting line is 0.432
The correlation coefficient is: 0.599
The mean square error is: 0.01815

Diatom 2011



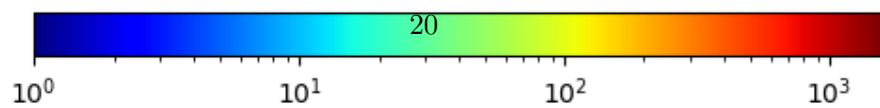
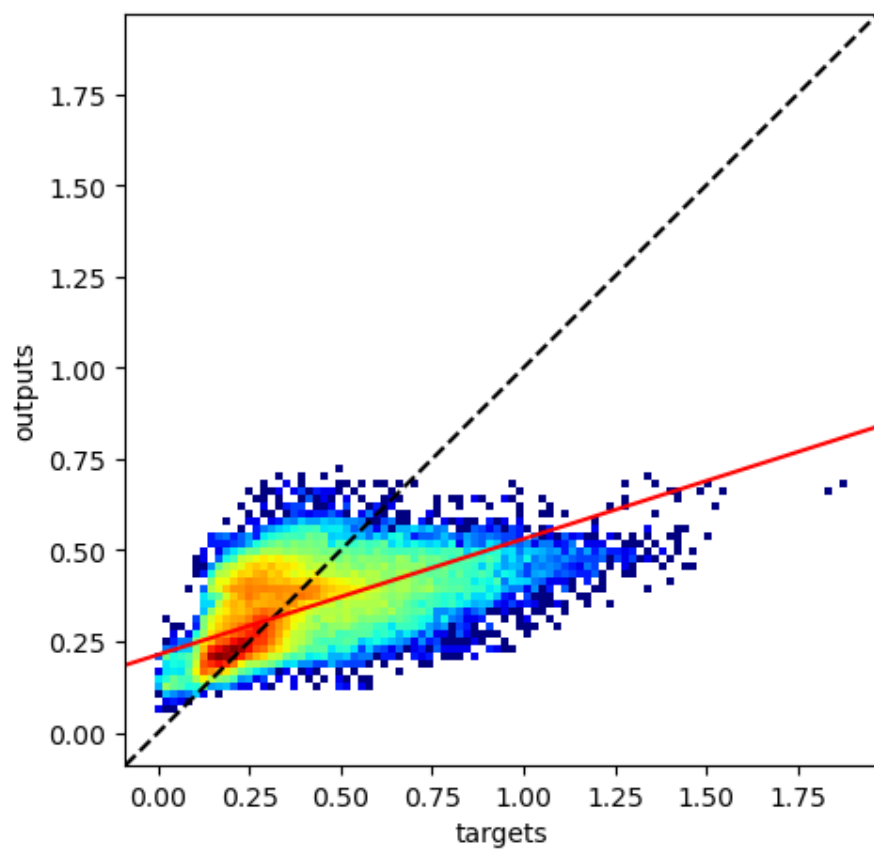
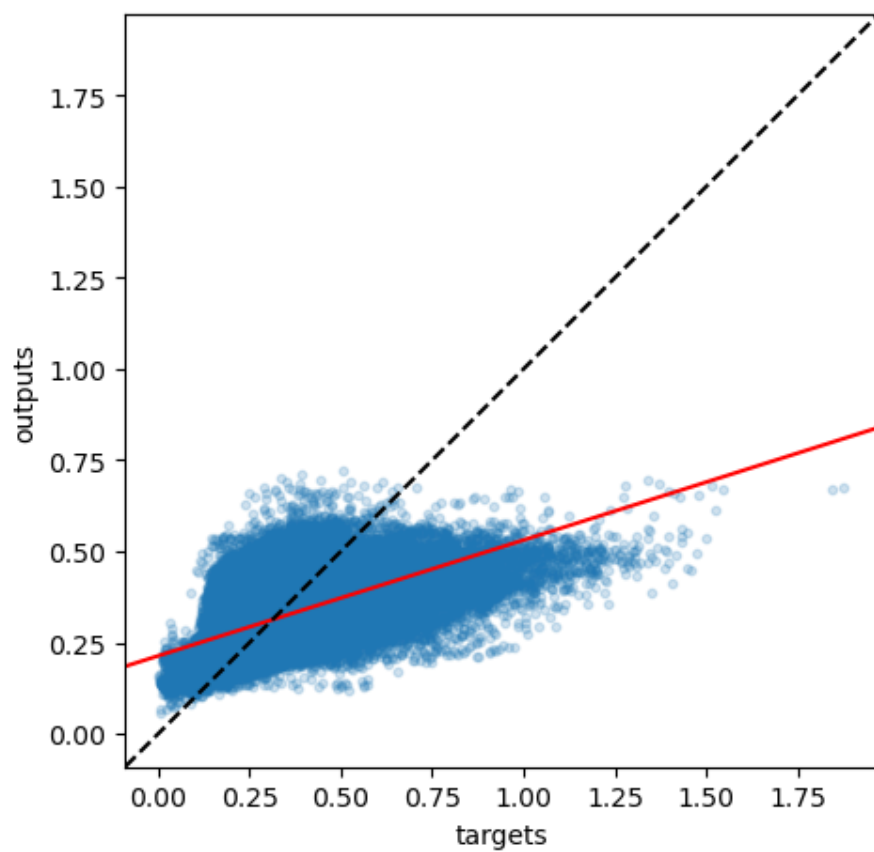
The amount of data points is 70794
The slope of the best fitting line is 0.45
The correlation coefficient is: 0.687
The mean square error is: 0.01352

Diatom 2012



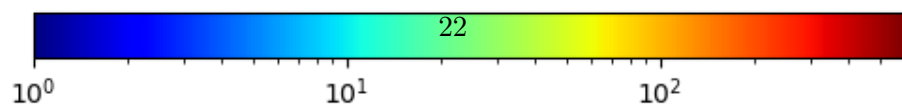
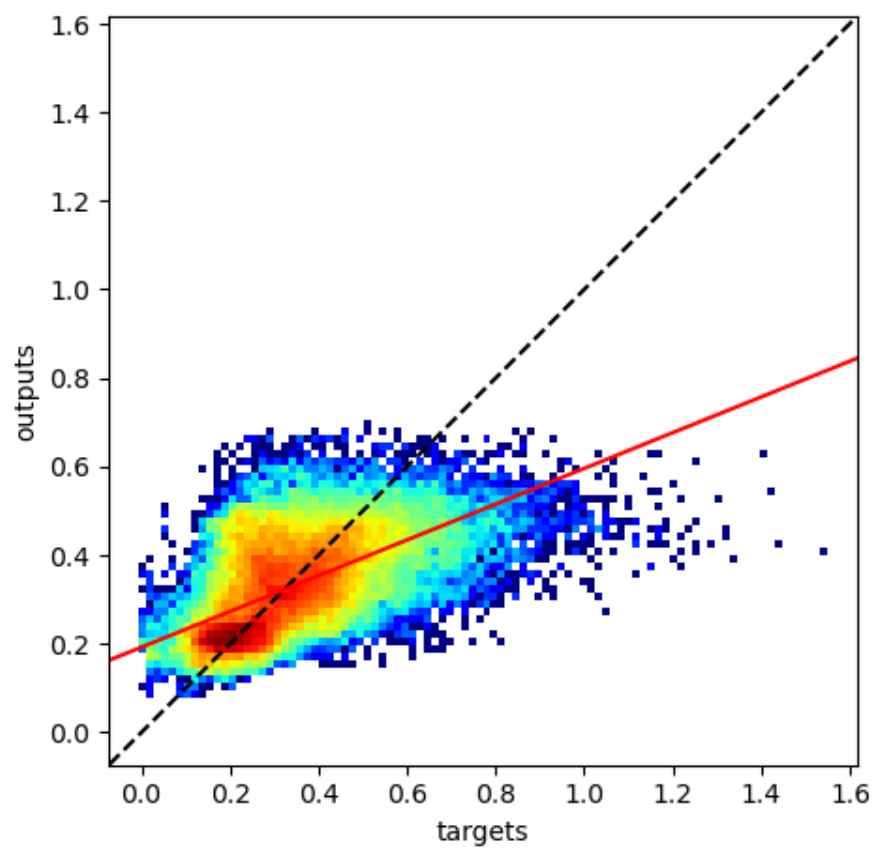
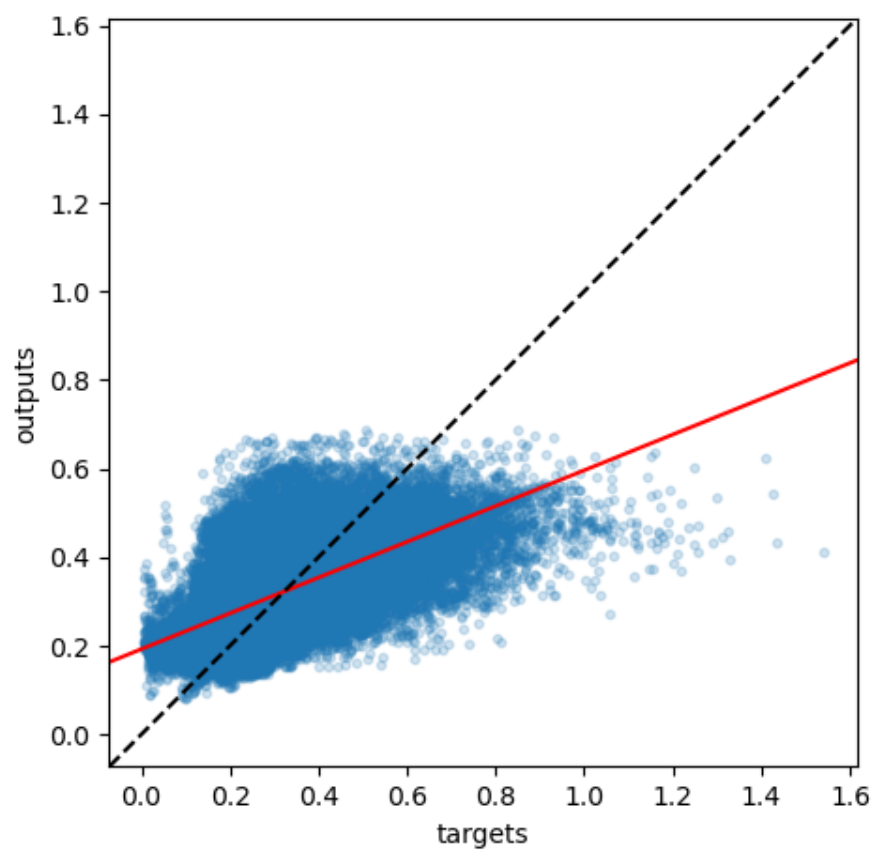
The amount of data points is 70794
The slope of the best fitting line is 0.317
The correlation coefficient is: 0.589
The mean square error is: 0.02121

Diatom 2013



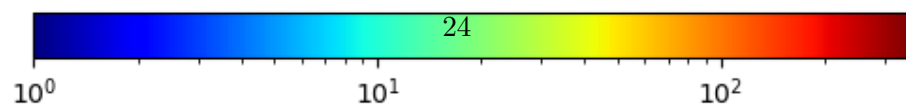
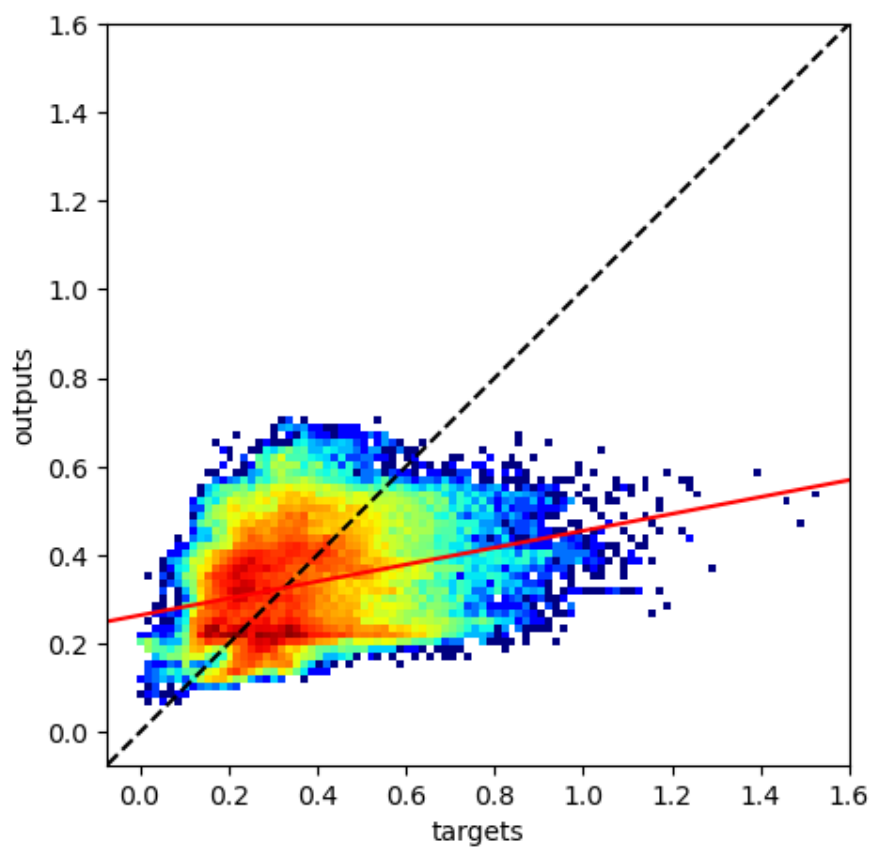
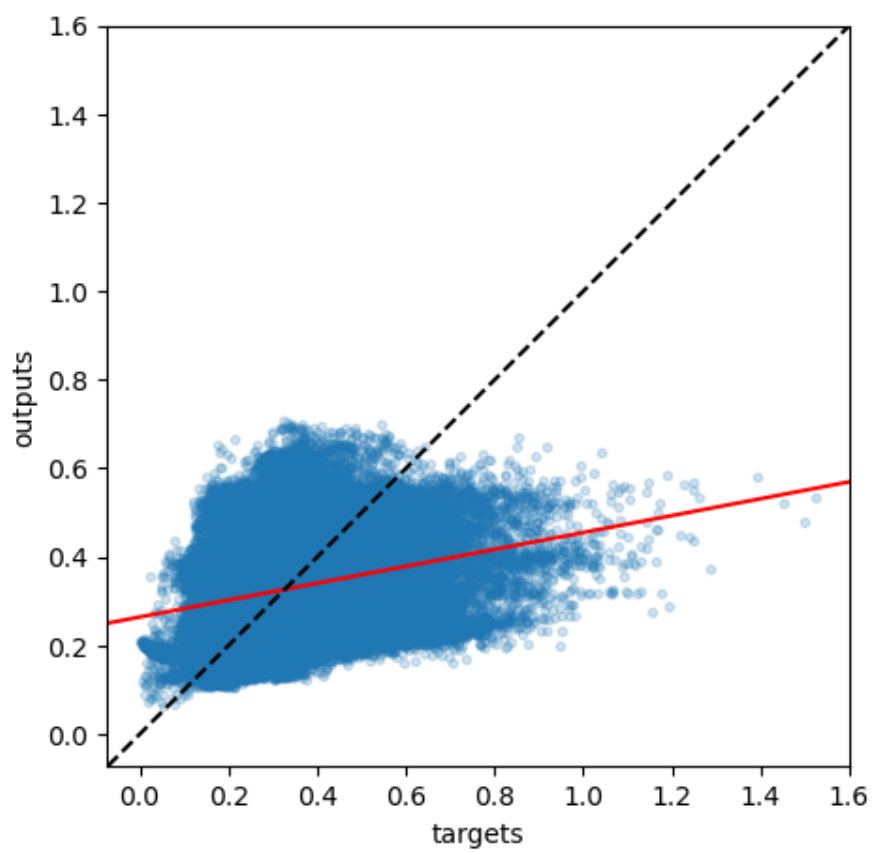
The amount of data points is 68931
The slope of the best fitting line is 0.404
The correlation coefficient is: 0.582
The mean square error is: 0.01414

Diatom 2014



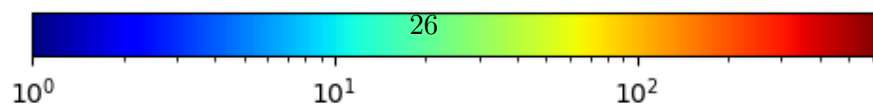
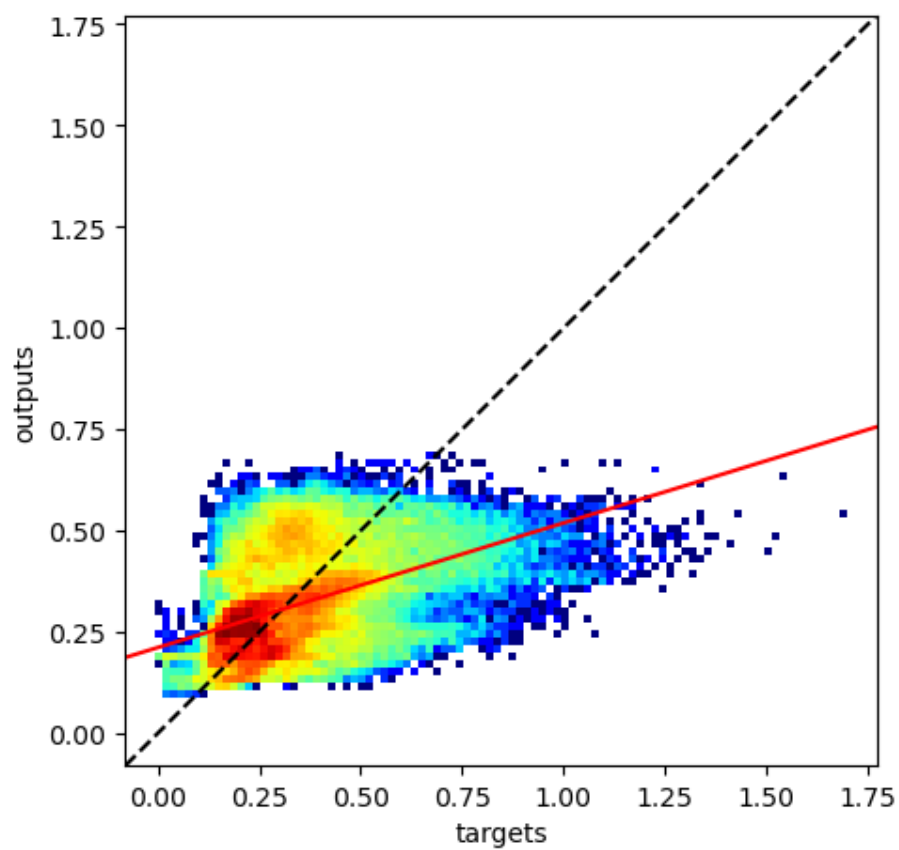
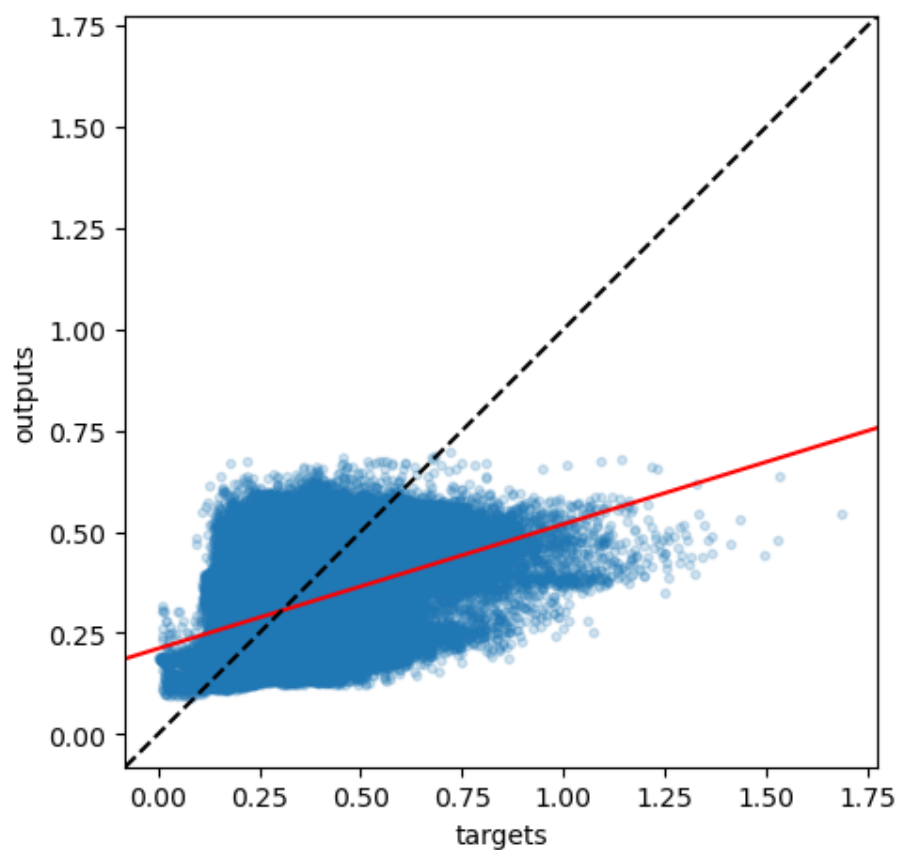
The amount of data points is 70794
The slope of the best fitting line is 0.191
The correlation coefficient is: 0.266
The mean square error is: 0.02648

Diatom 2015



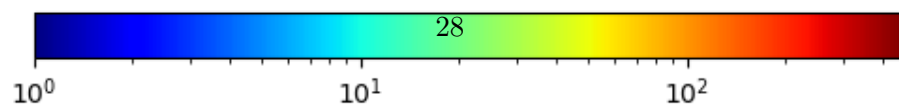
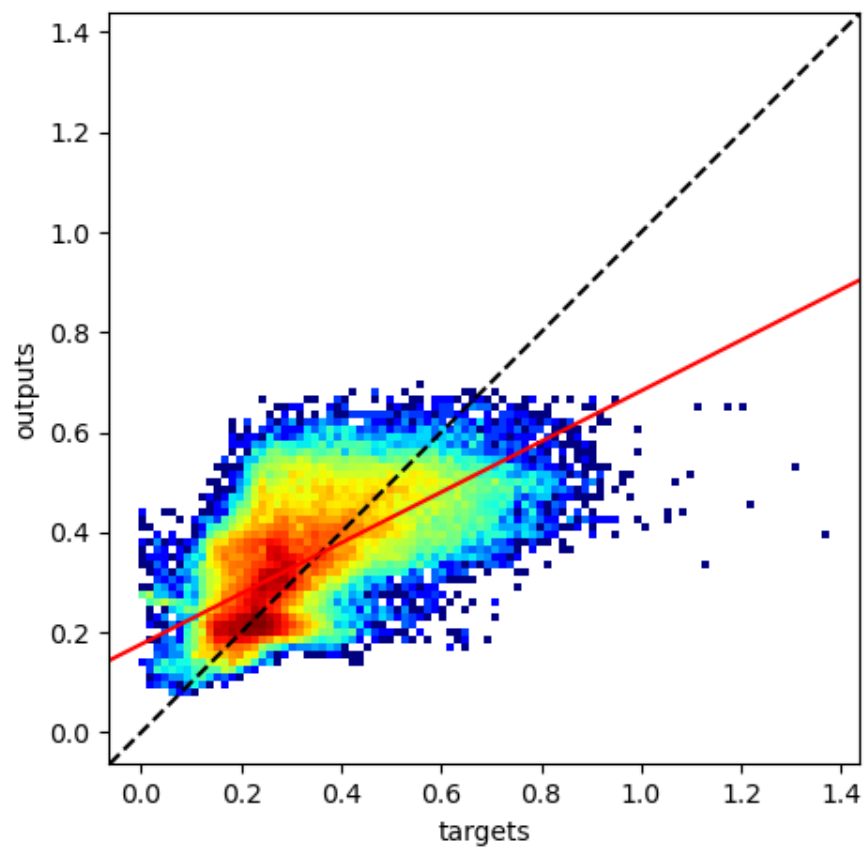
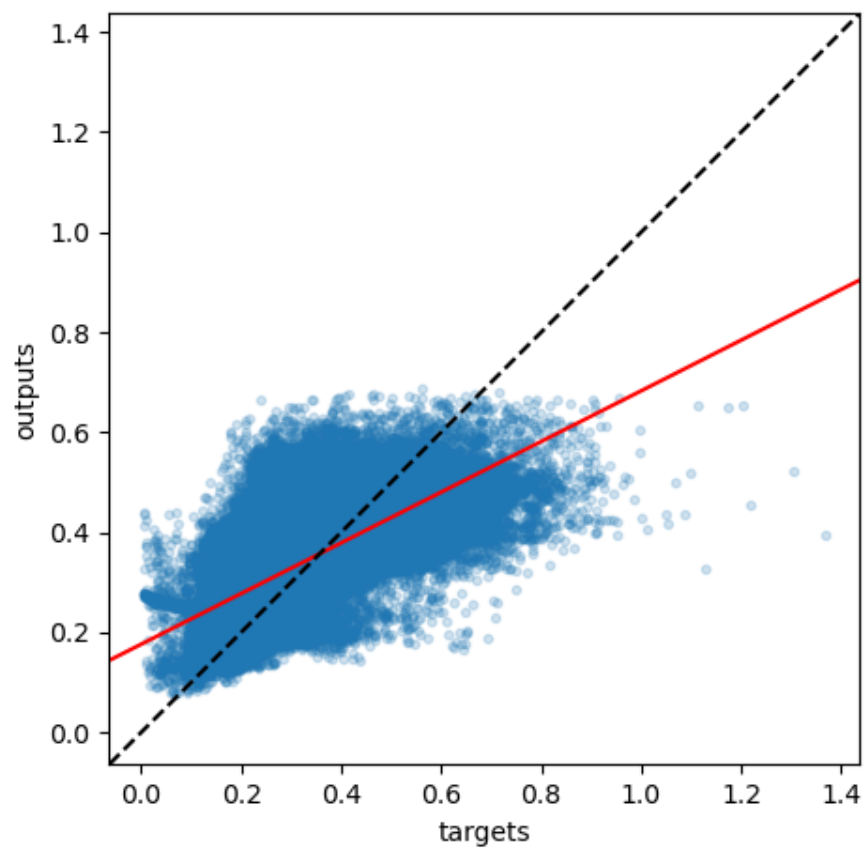
The amount of data points is 70794
The slope of the best fitting line is 0.307
The correlation coefficient is: 0.455
The mean square error is: 0.02429

Diatom 2016



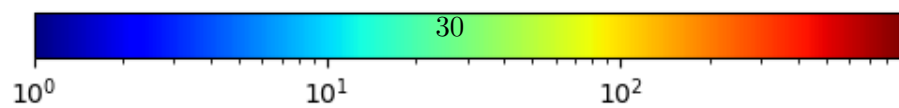
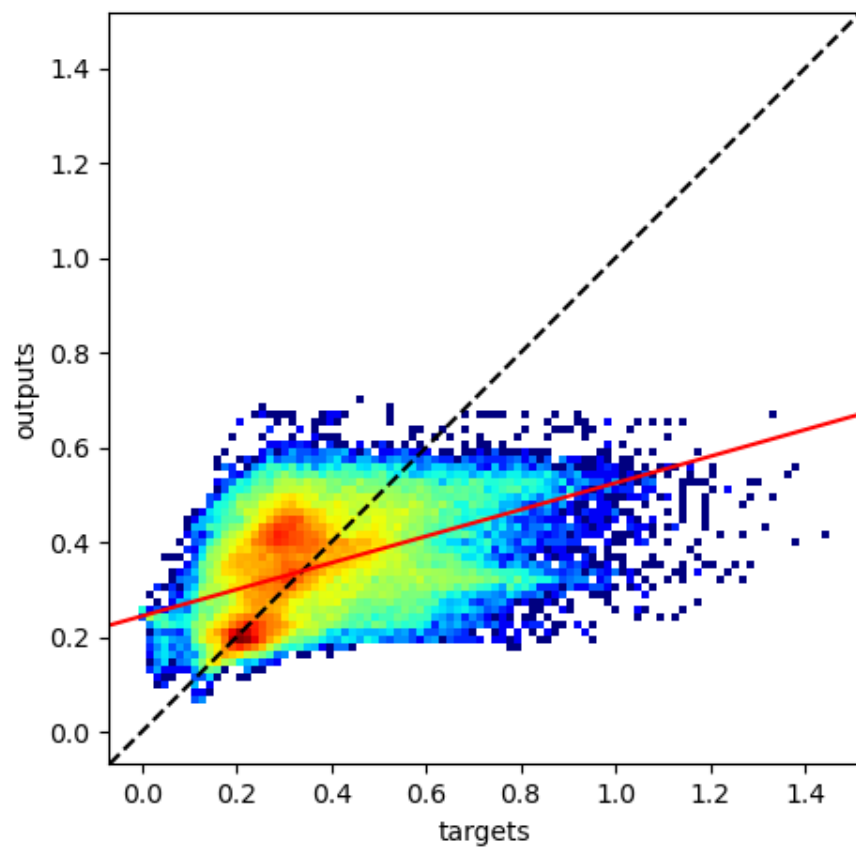
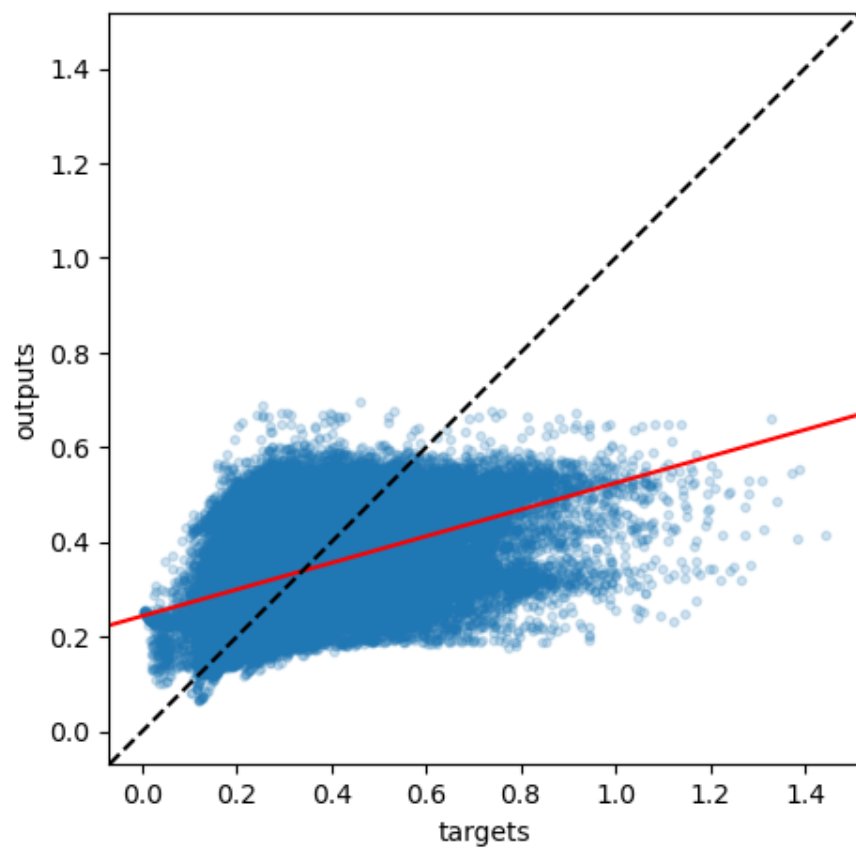
The amount of data points is 68931
The slope of the best fitting line is 0.507
The correlation coefficient is: 0.633
The mean square error is: 0.01183

Diatom 2017



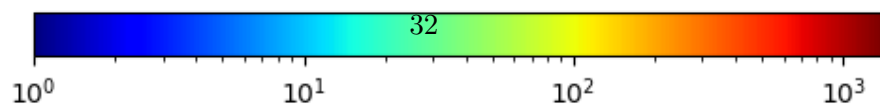
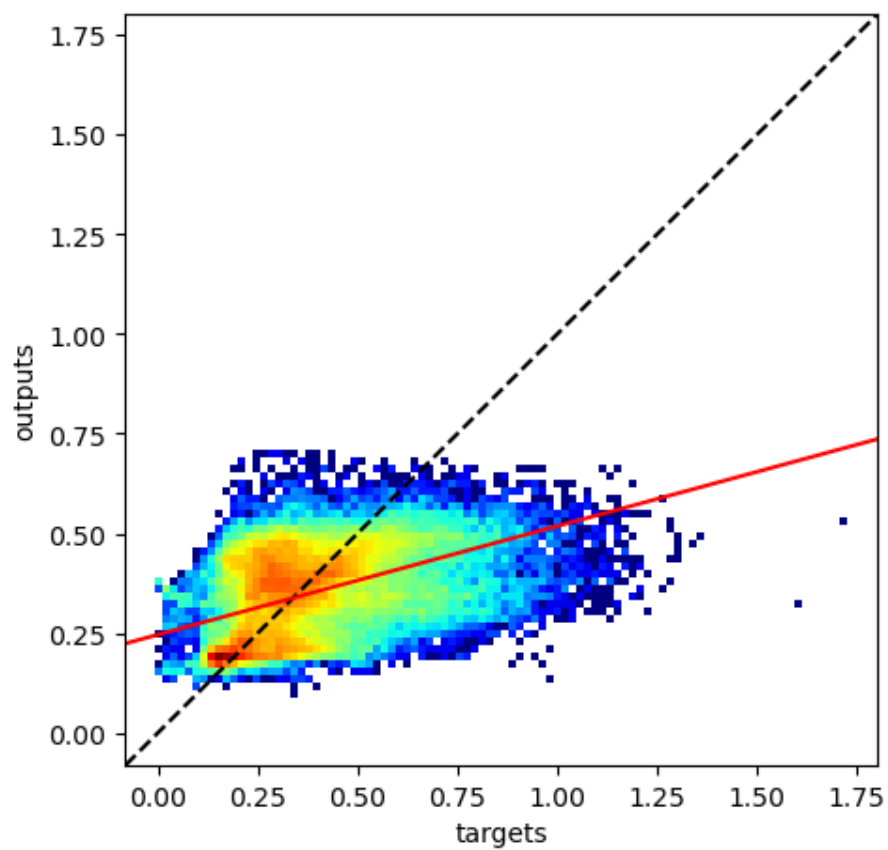
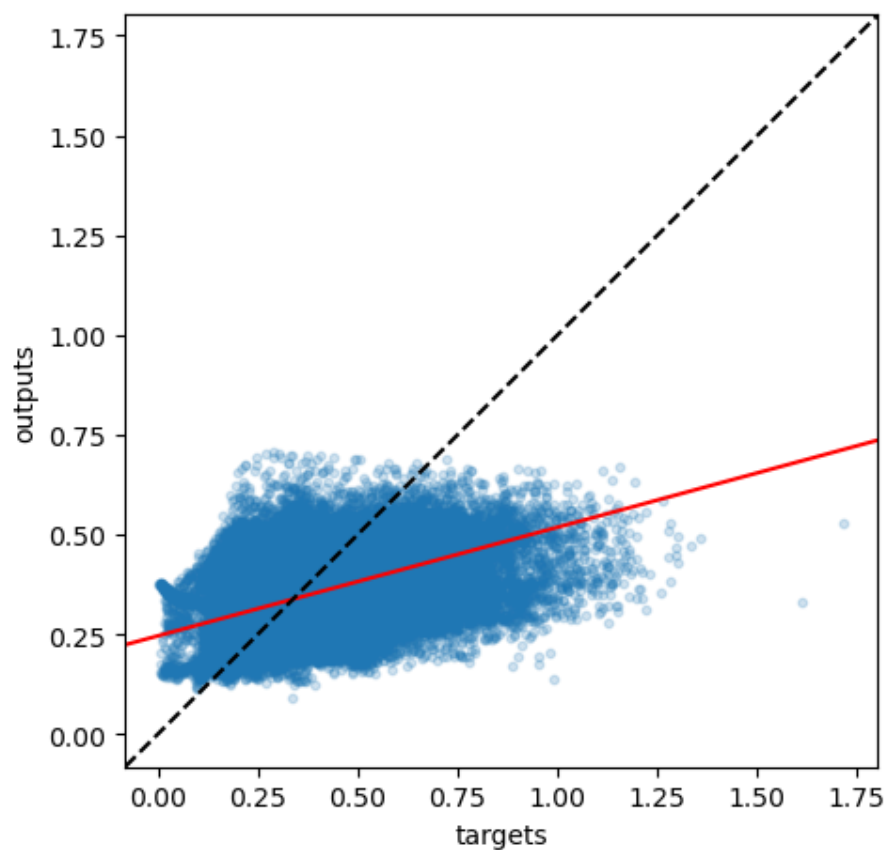
The amount of data points is 70794
The slope of the best fitting line is 0.281
The correlation coefficient is: 0.435
The mean square error is: 0.02128

Diatom 2018



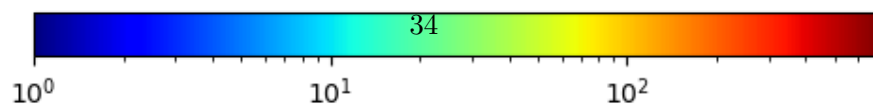
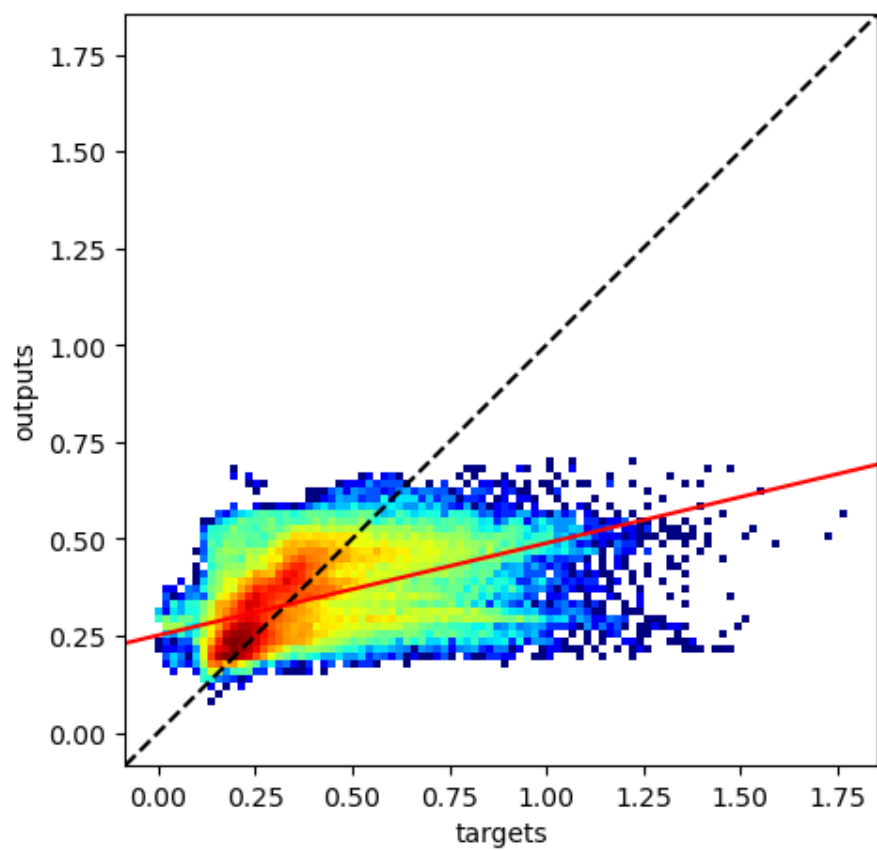
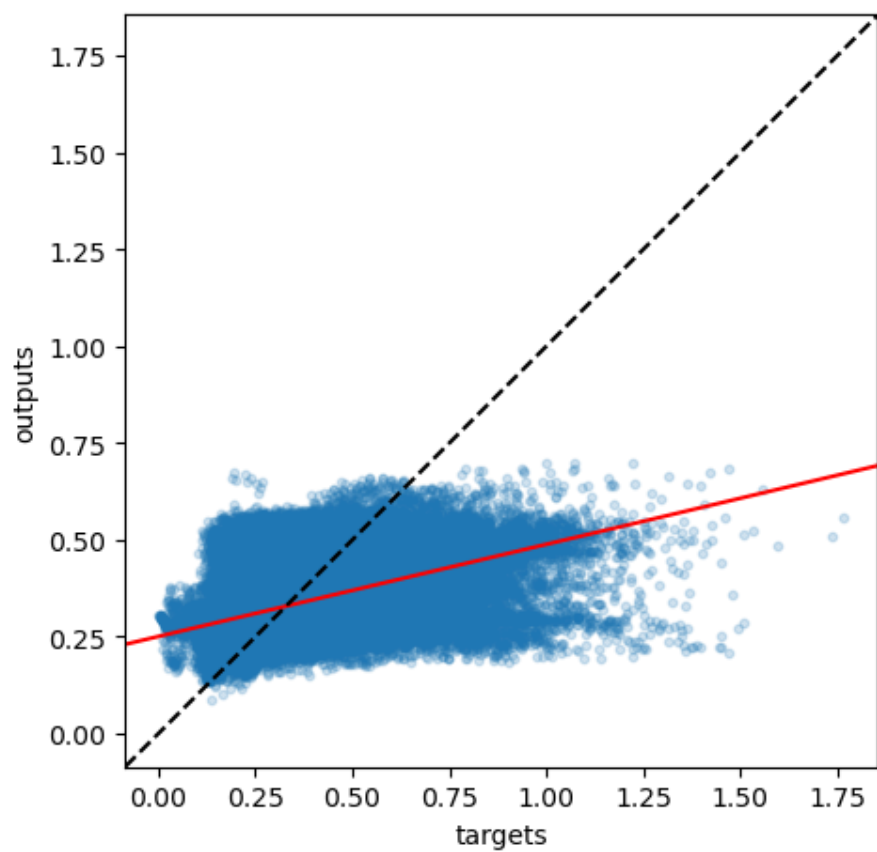
The amount of data points is 68931
The slope of the best fitting line is 0.272
The correlation coefficient is: 0.451
The mean square error is: 0.02484

Diatom 2019



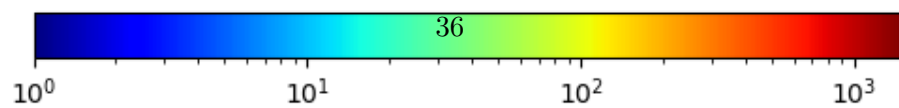
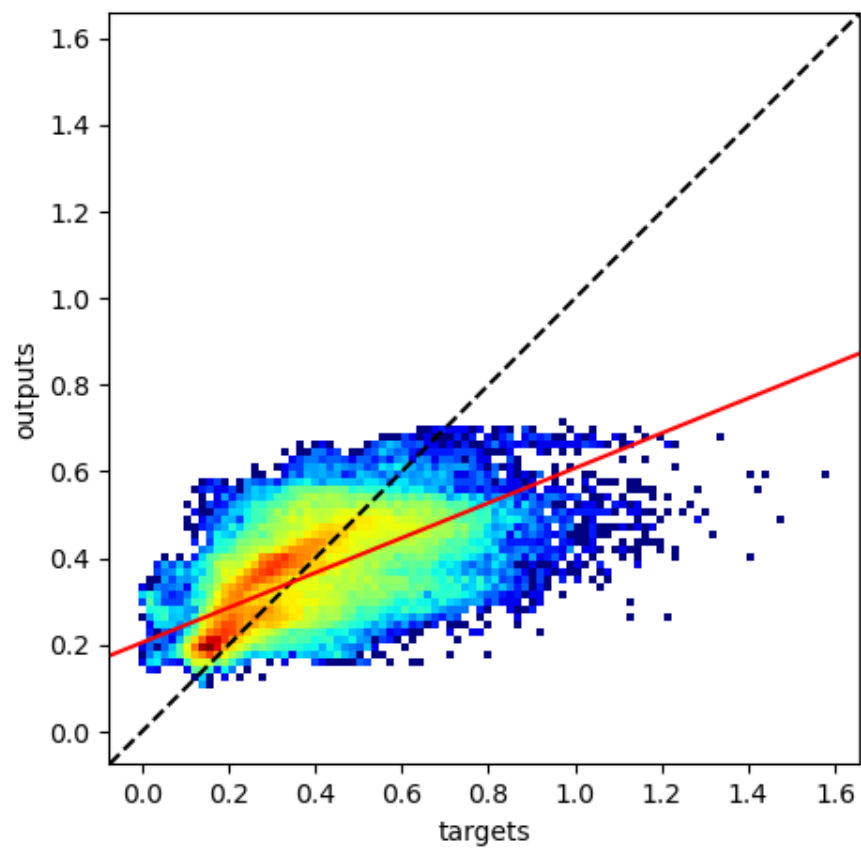
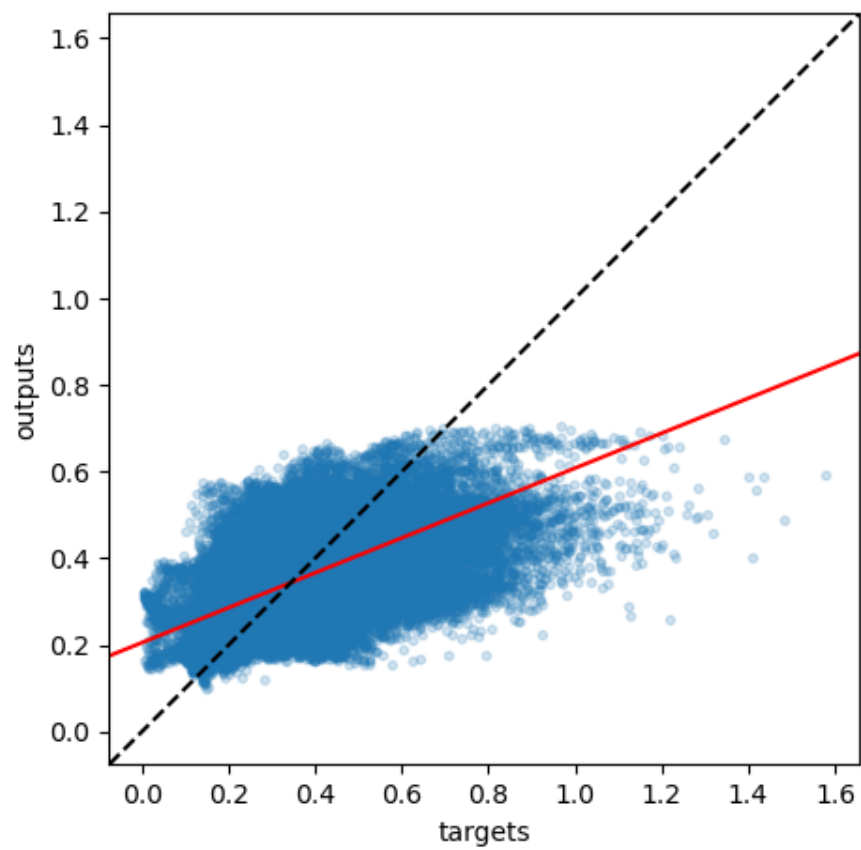
The amount of data points is 70794
The slope of the best fitting line is 0.238
The correlation coefficient is: 0.484
The mean square error is: 0.03245

Diatom 2020



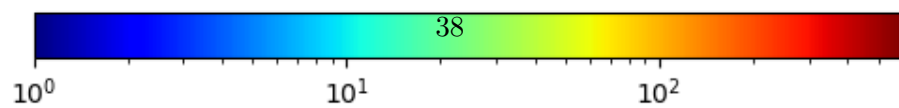
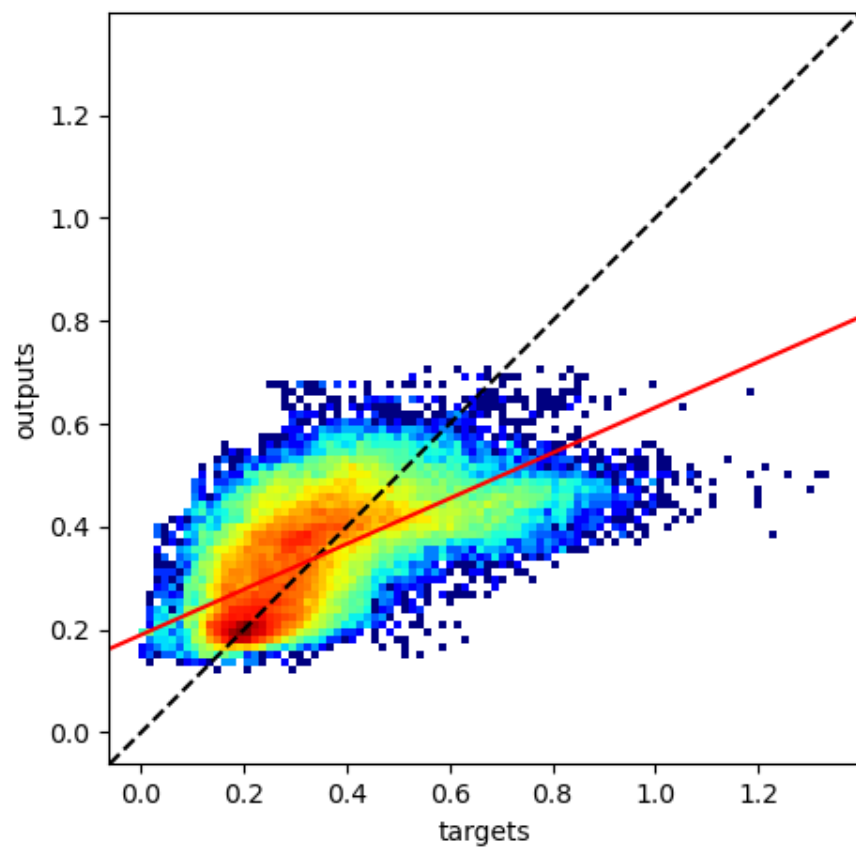
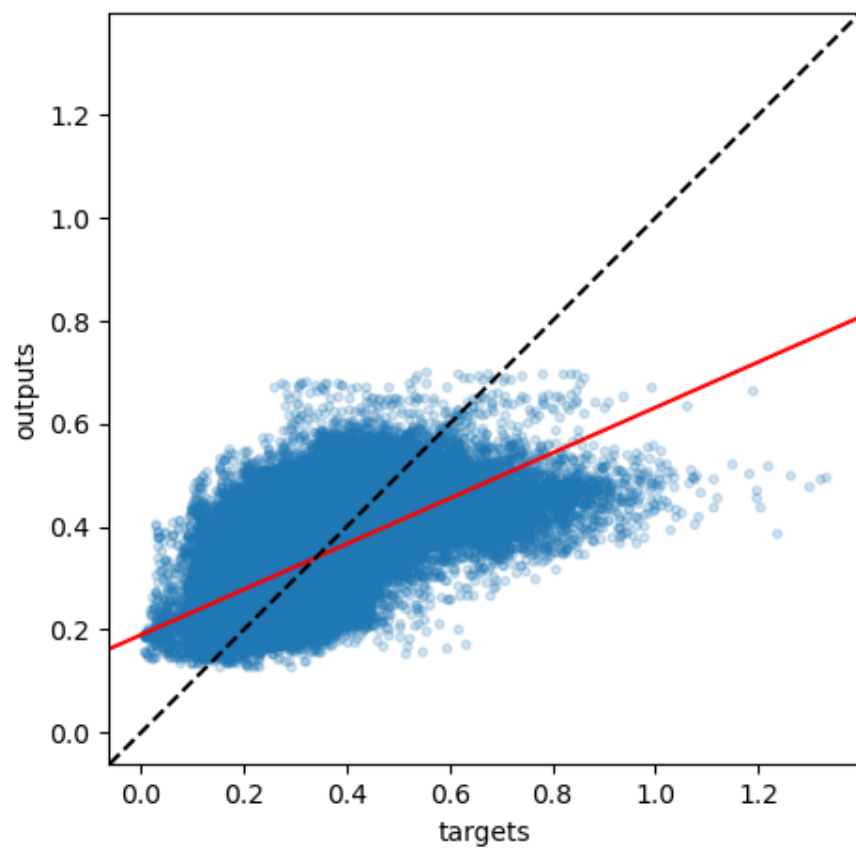
The amount of data points is 70794
The slope of the best fitting line is 0.403
The correlation coefficient is: 0.653
The mean square error is: 0.0173

Diatom 2021



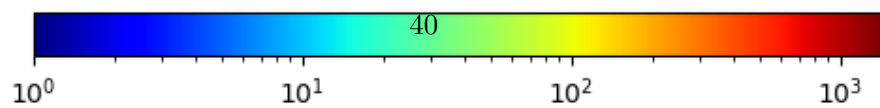
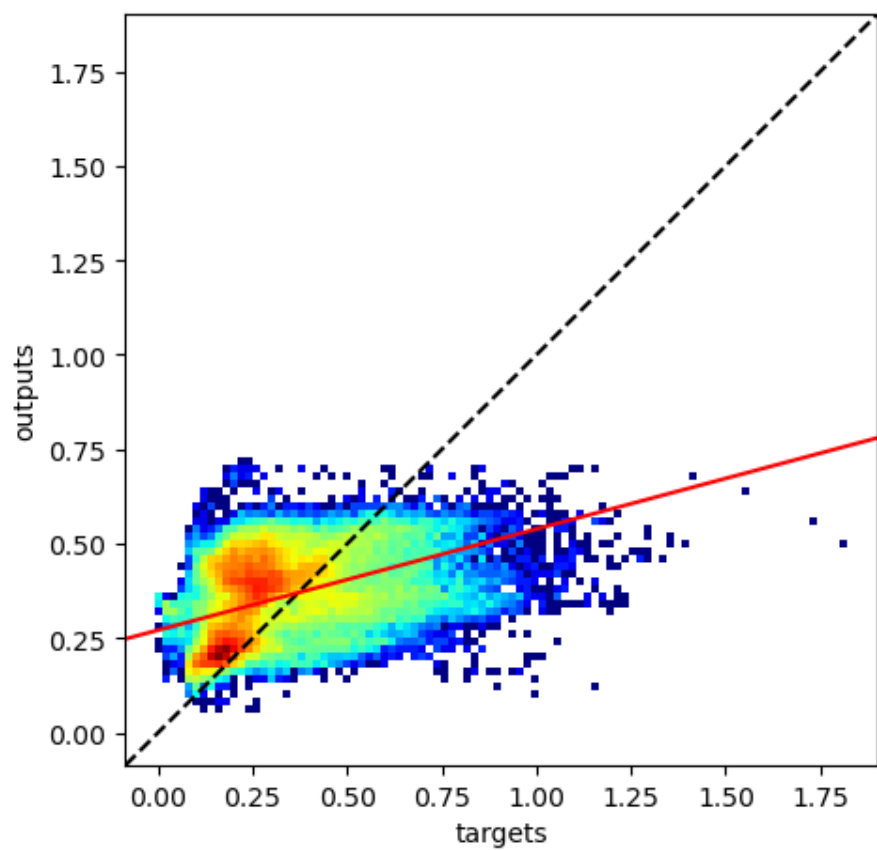
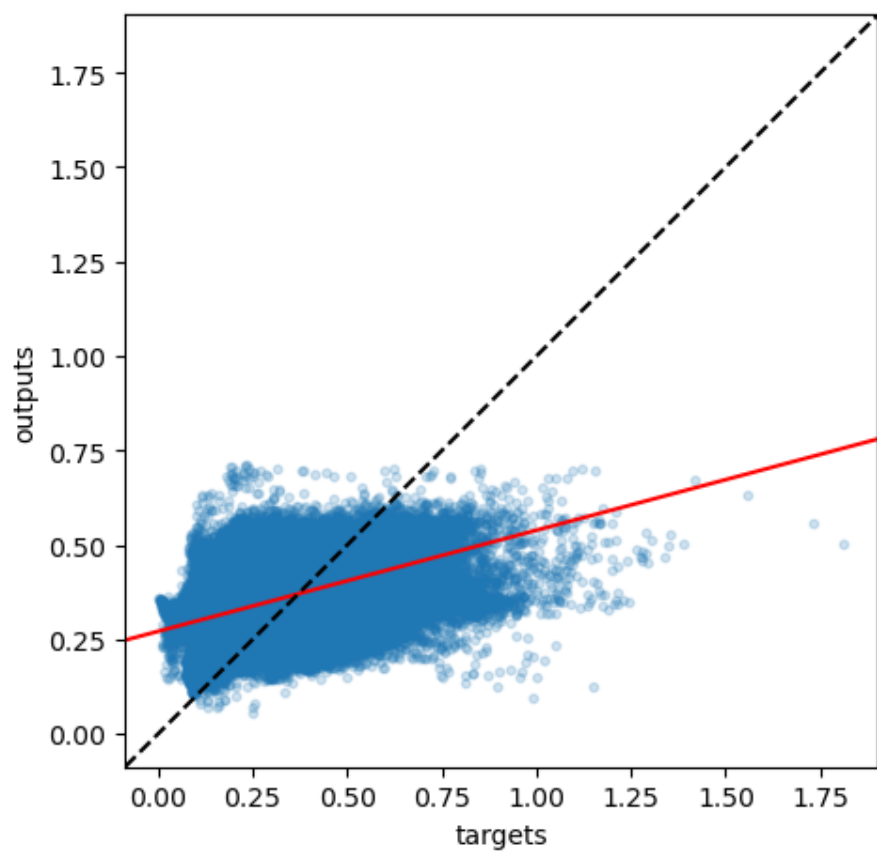
The amount of data points is 68931
The slope of the best fitting line is 0.442
The correlation coefficient is: 0.629
The mean square error is: 0.01305

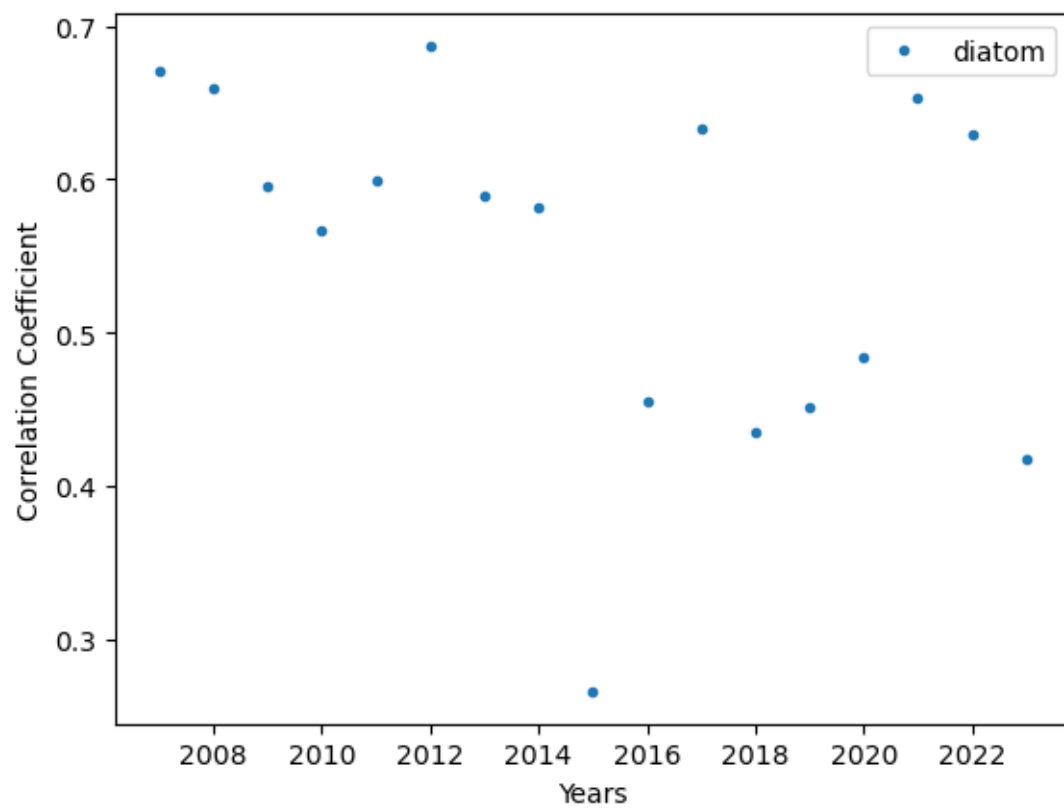
Diatom 2022

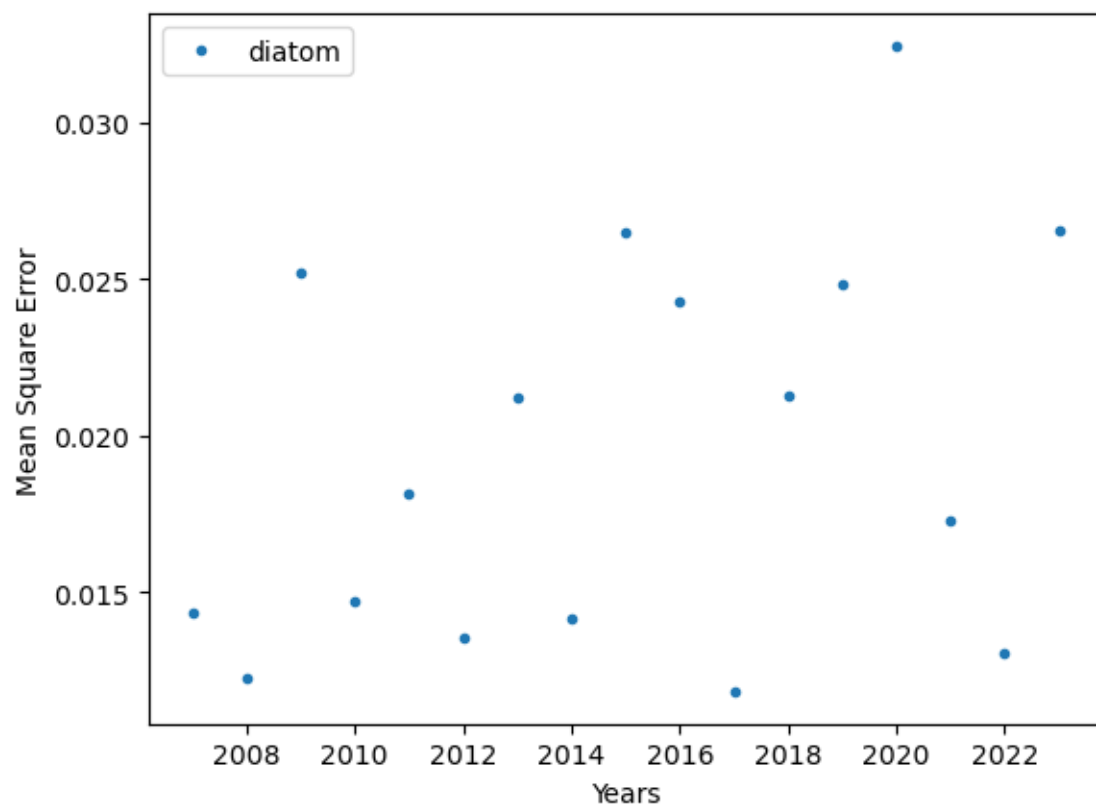


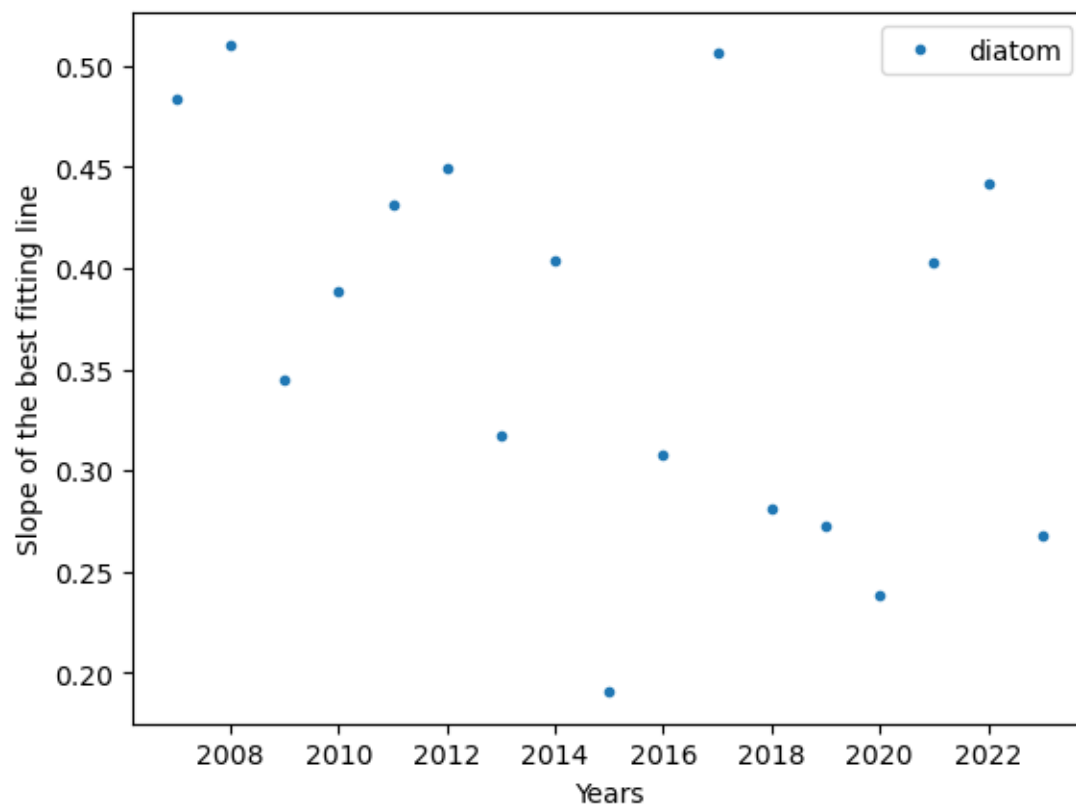
The amount of data points is 70794
The slope of the best fitting line is 0.268
The correlation coefficient is: 0.417
The mean square error is: 0.02655

Diatom 2023









1.7 Other Years (Daily)

```
[ ]: r_all2 = np.array([])
rms_all2 = np.array([])
slope_all2 = np.array([])

for i in tqdm(range(0, len(ds.time_counter))):

    dataset = ds.isel(time_counter=i)

    drivers, diat, _ = datasets_preparation(dataset)

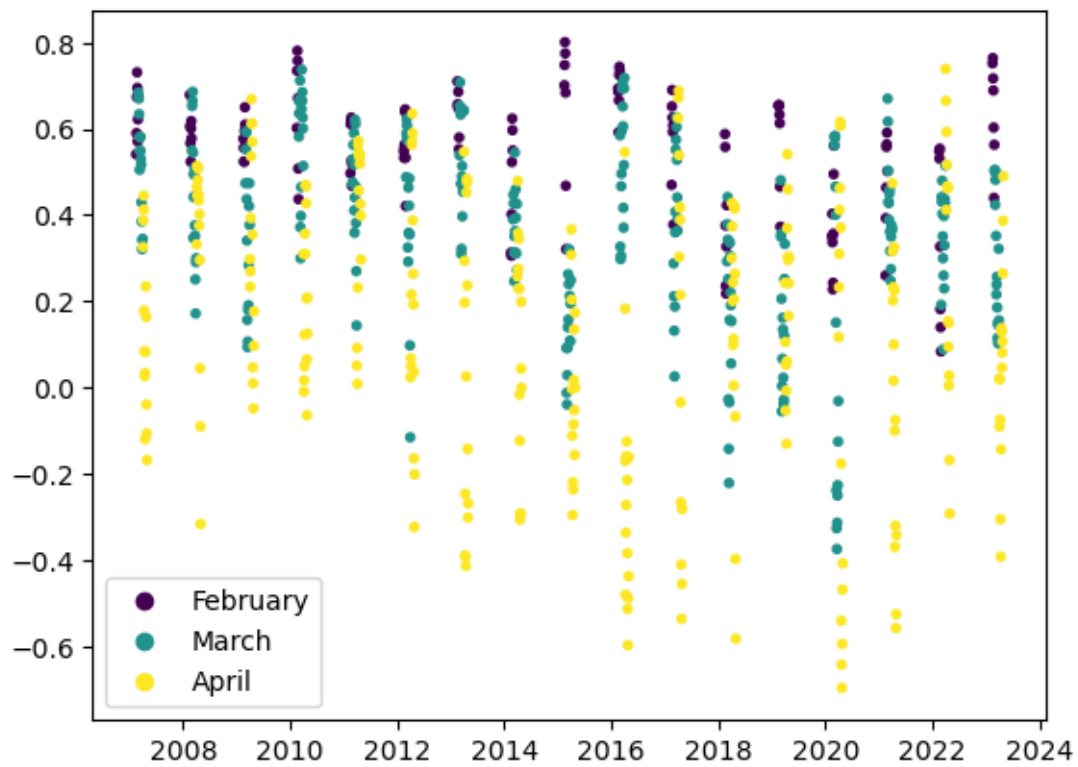
    r, rms, m = regressor3(drivers, diat)

    r_all2 = np.append(r_all2,r)
    rms_all2 = np.append(rms_all2,rms)
    slope_all2 = np.append(slope_all2,m)

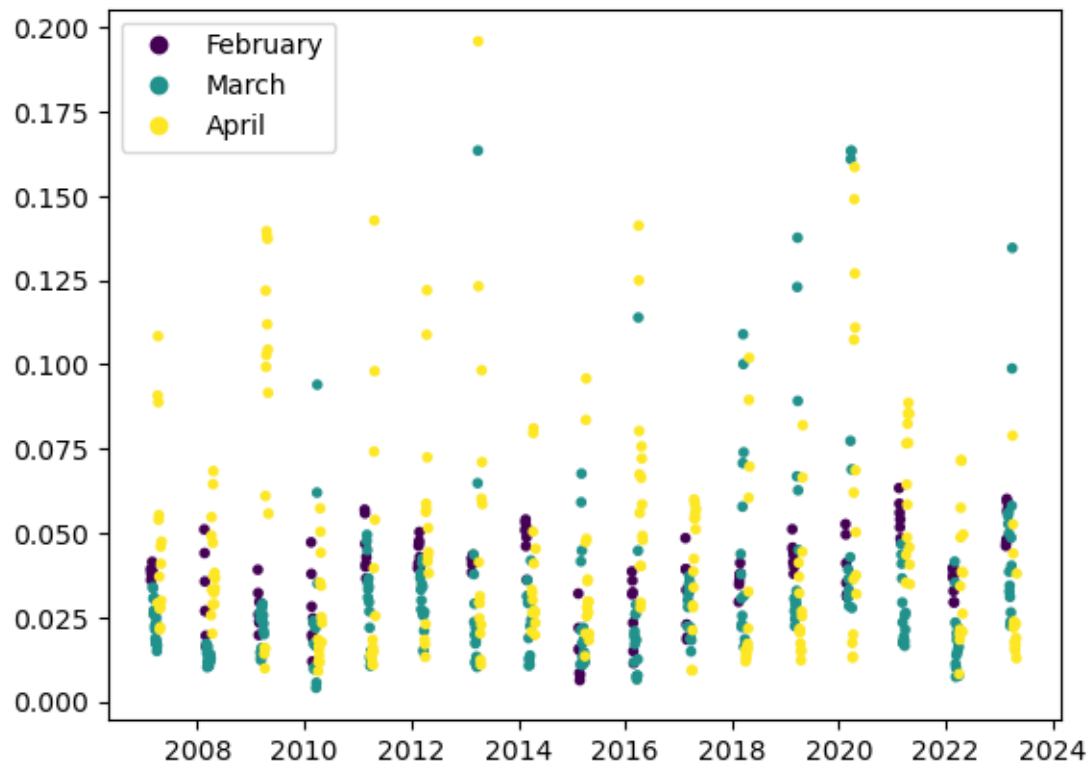
plotting2(r_all2, 'Correlation Coefficients')
plotting2(rms_all2, 'Mean Square Errors')
plotting2(slope_all2, 'Slope of the best fitting line')
```

0%| | 0/640 [00:00<?, ?it/s]

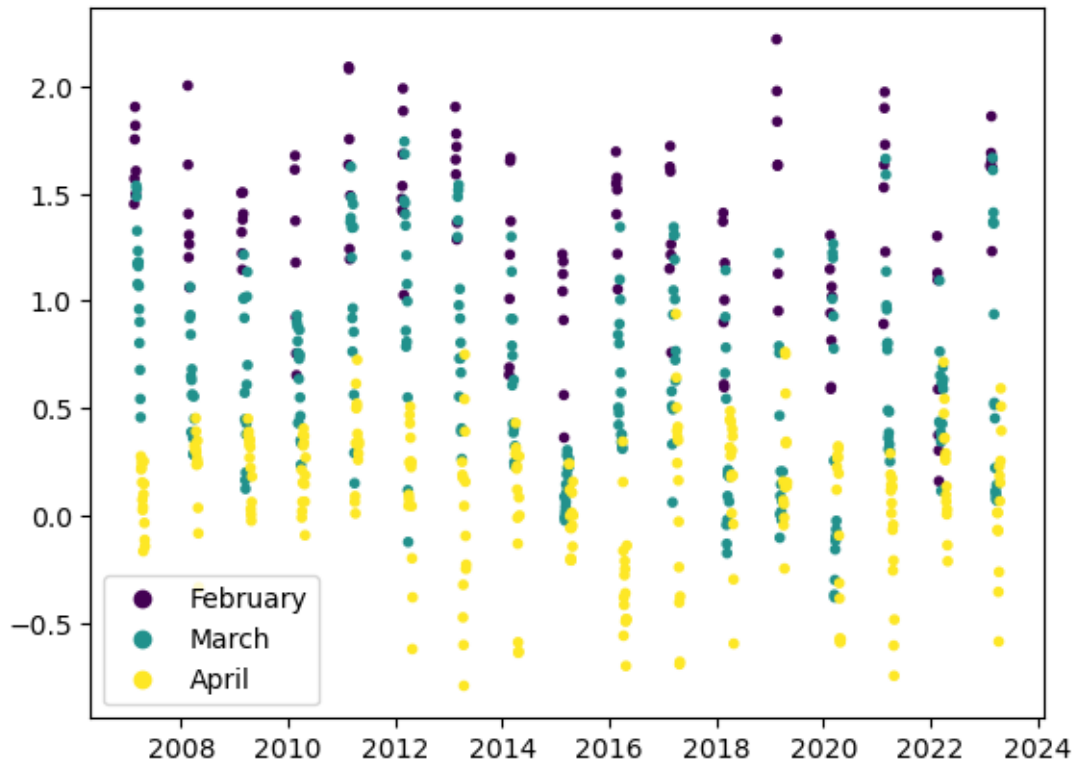
Daily Correlation Coefficients (15 Feb - 30 Apr)



Daily Mean Square Errors (15 Feb - 30 Apr)



Daily Slope of the best fitting line (15 Feb - 30 Apr)



2 Daily Maps

```
[ ]: maps = random.sample(range(0,len(ds.time_counter)),10)

for i in tqdm(maps):

    dataset = ds.isel(time_counter=i)
    drivers, diat, indx = datasets_preparation(dataset)

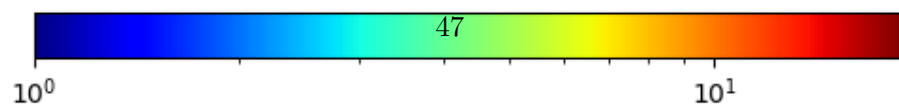
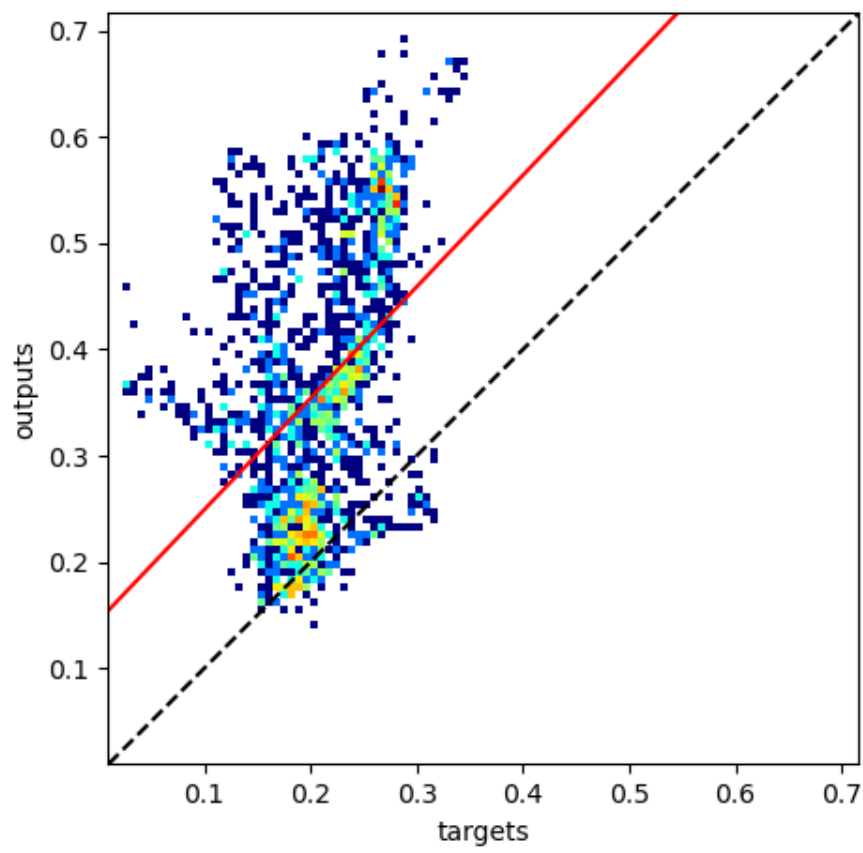
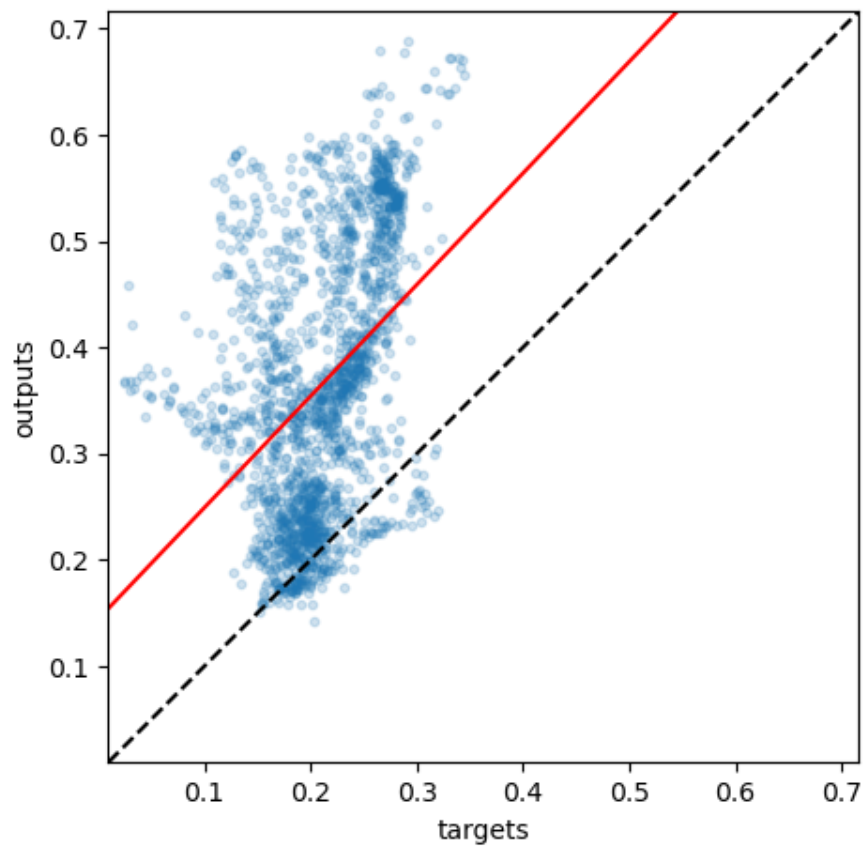
    diat_i = dataset['Diatom']

    regressor4(drivers, diat, 'Diatom ')
```

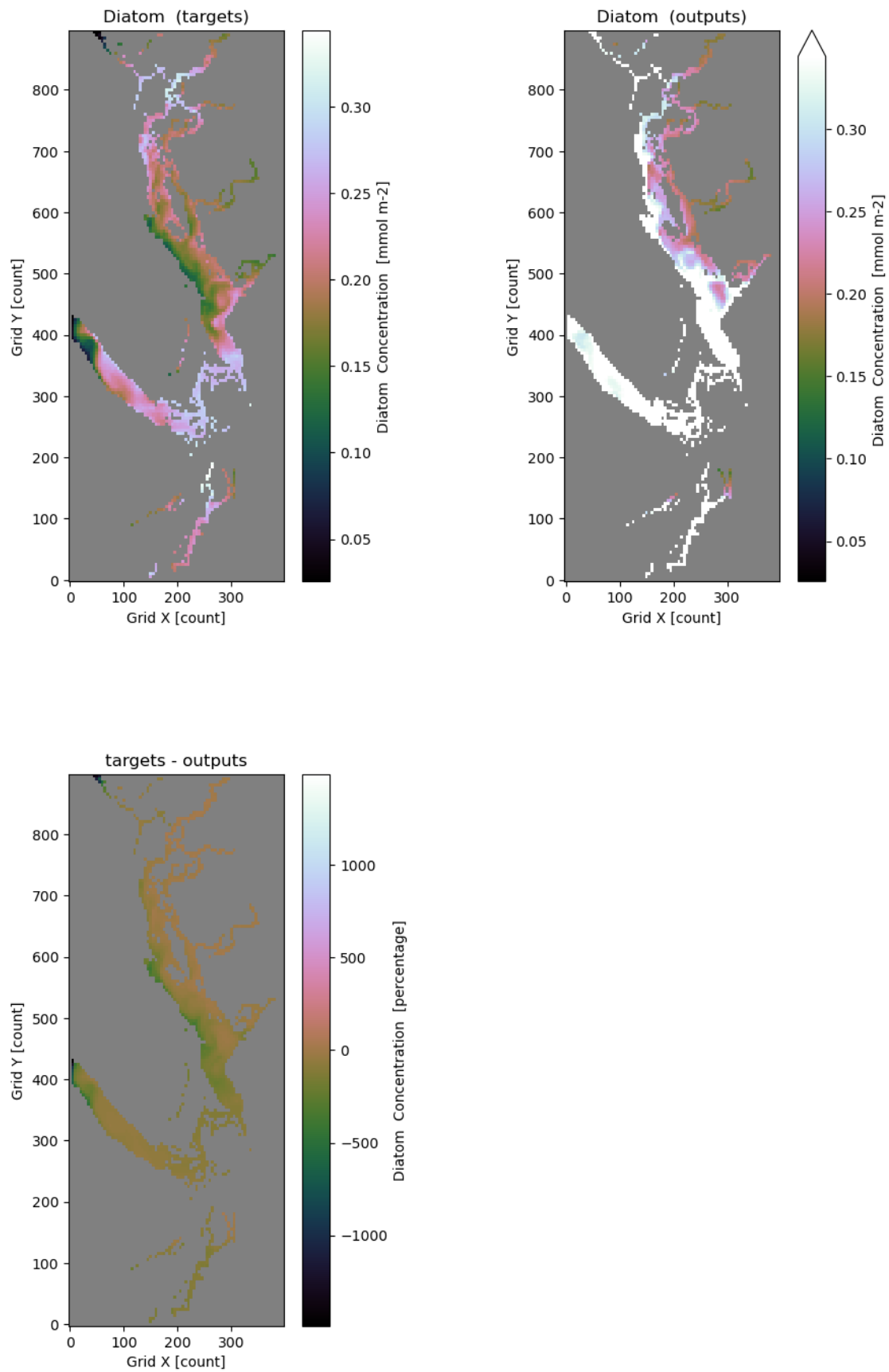
0%| | 0/10 [00:00<?, ?it/s]

The amount of data points is 1863
 The slope of the best fitting line is 1.049
 The correlation coefficient is: 0.408
 The mean square error is: 0.03732

Diatom 2017-03-14

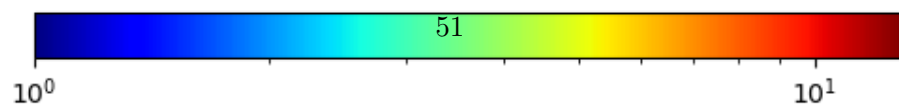
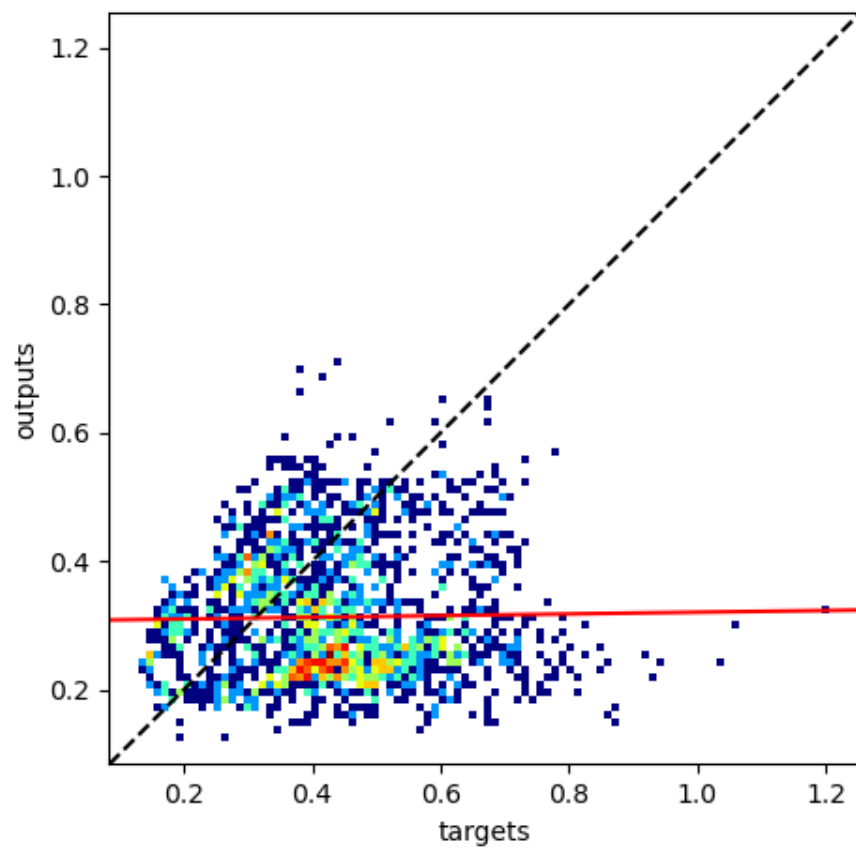
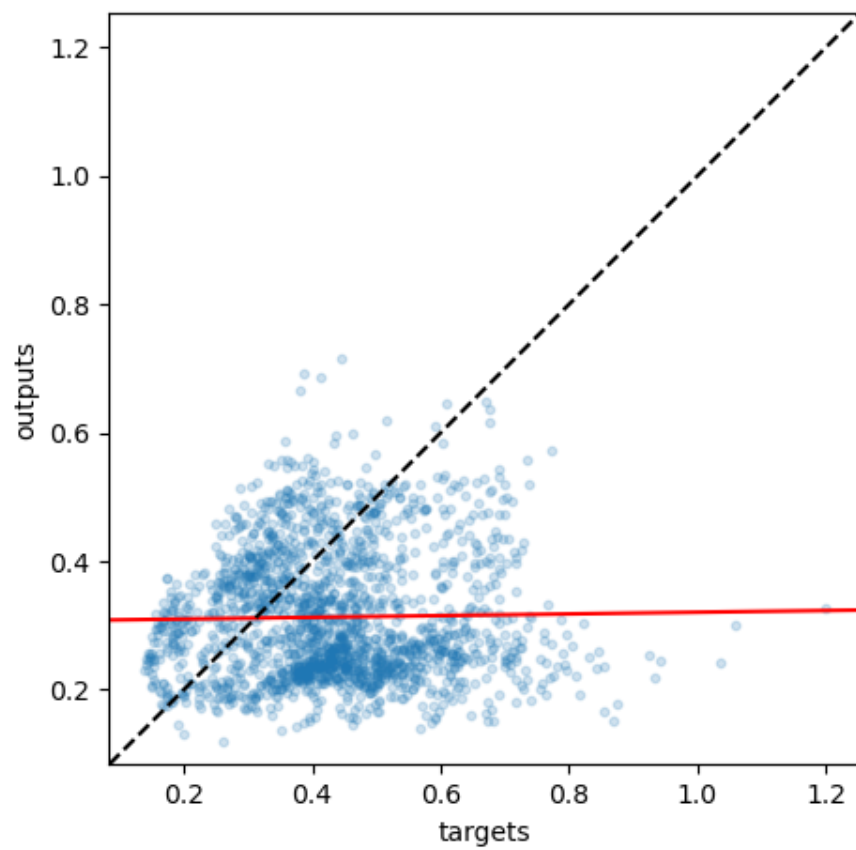


2017-03-14

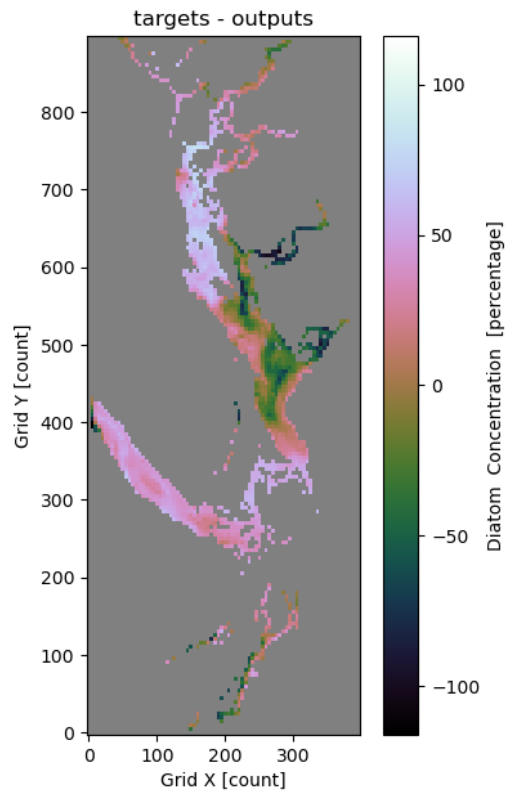
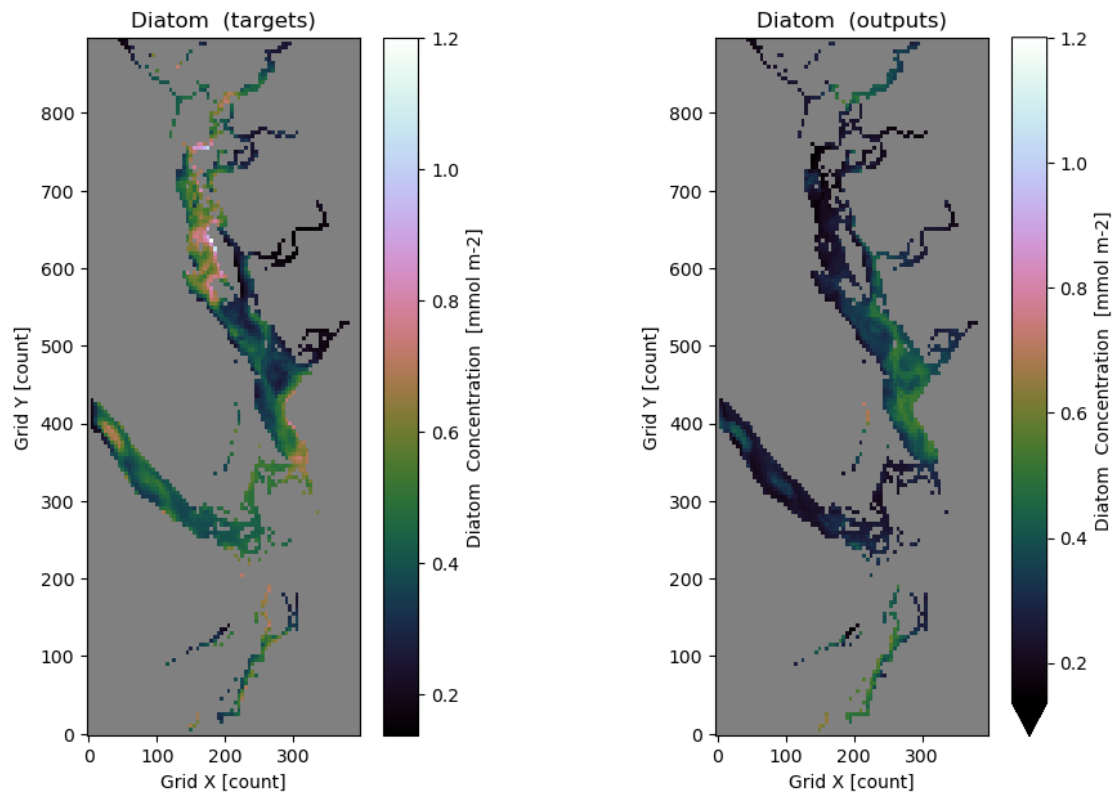


The amount of data points is 1863
The slope of the best fitting line is 0.013
The correlation coefficient is: 0.019
The mean square error is: 0.04392

Diatom 2023-04-06

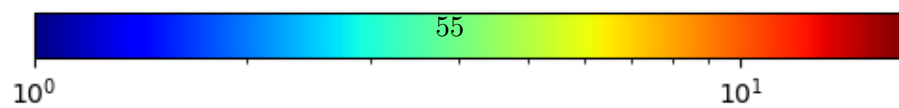
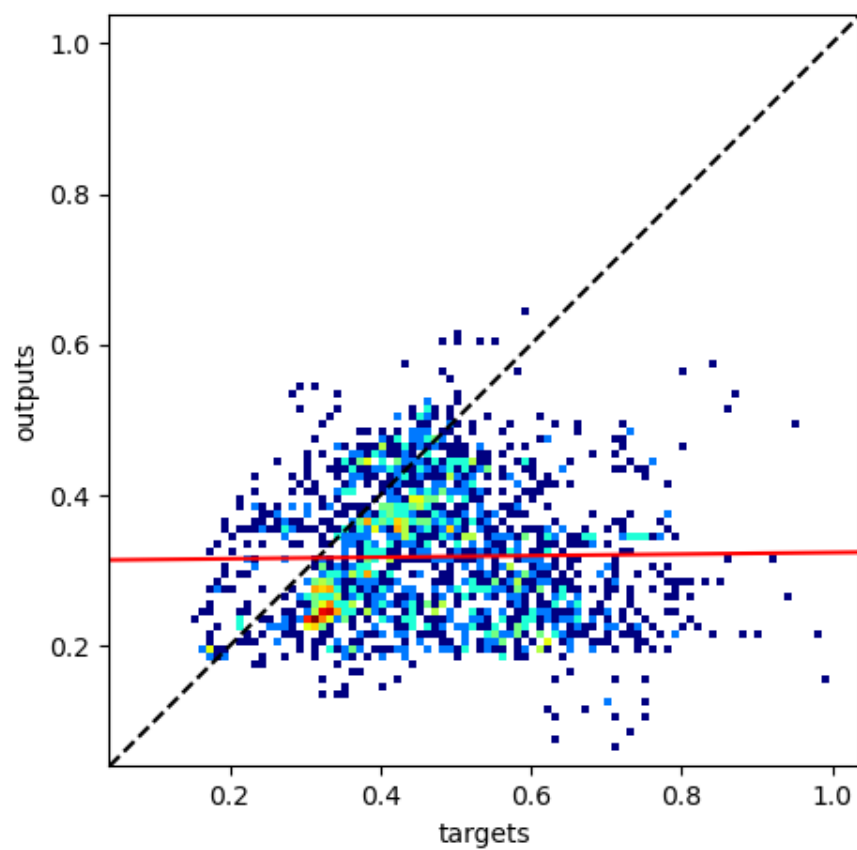
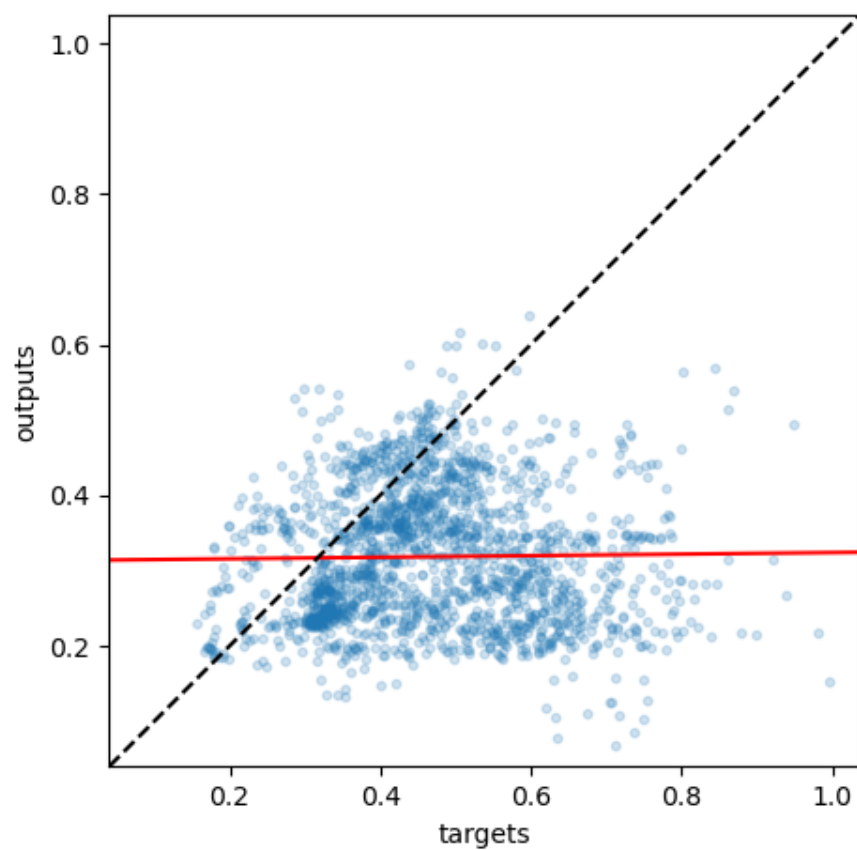


2023-04-06

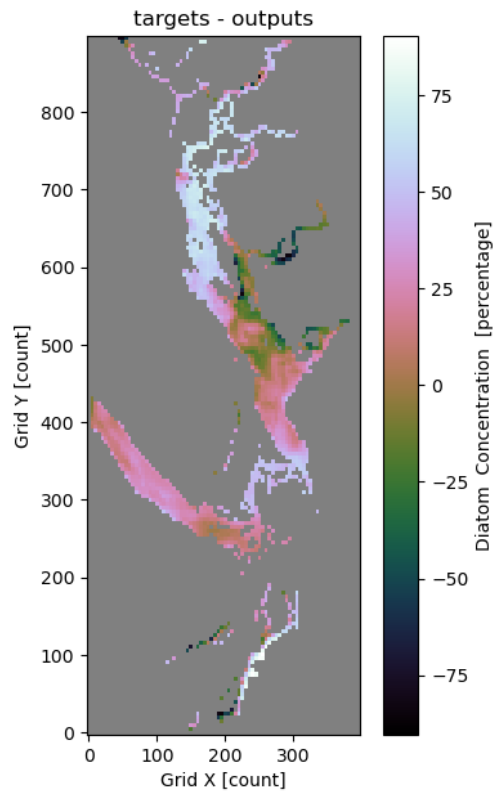
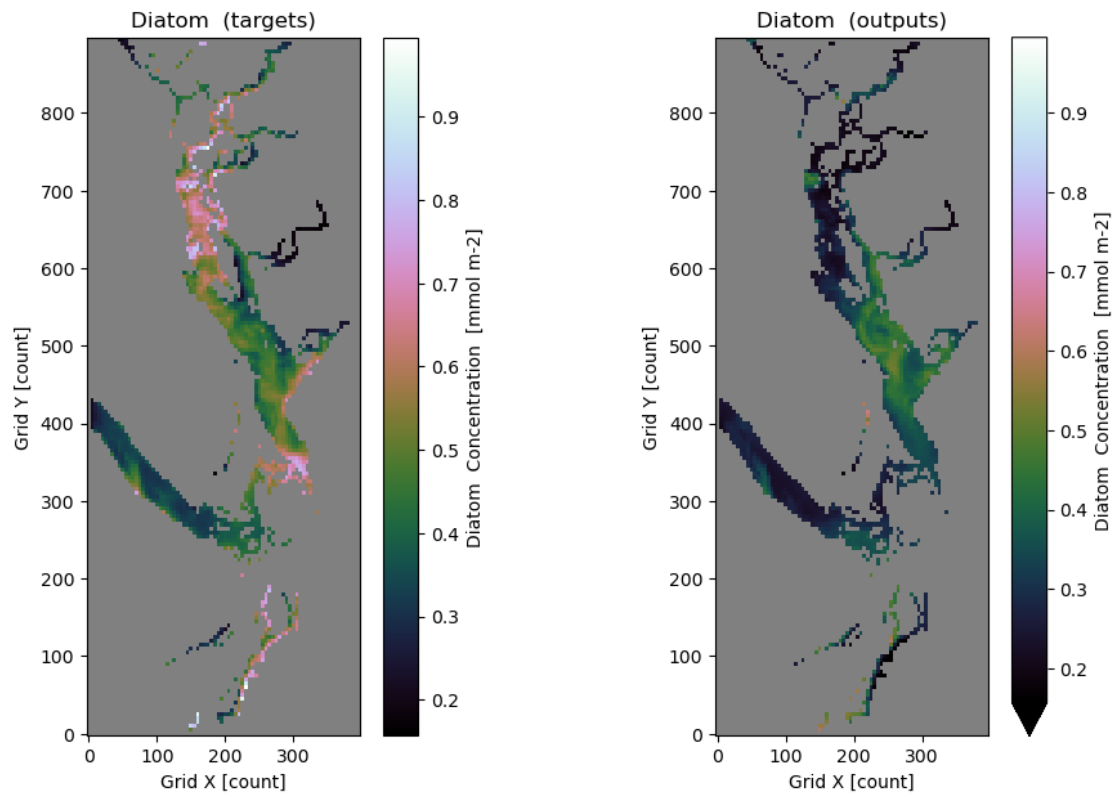


The amount of data points is 1863
The slope of the best fitting line is 0.01
The correlation coefficient is: 0.016
The mean square error is: 0.04869

Diatom 2021-04-10

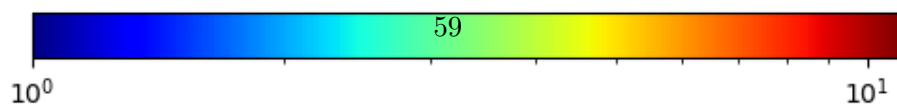
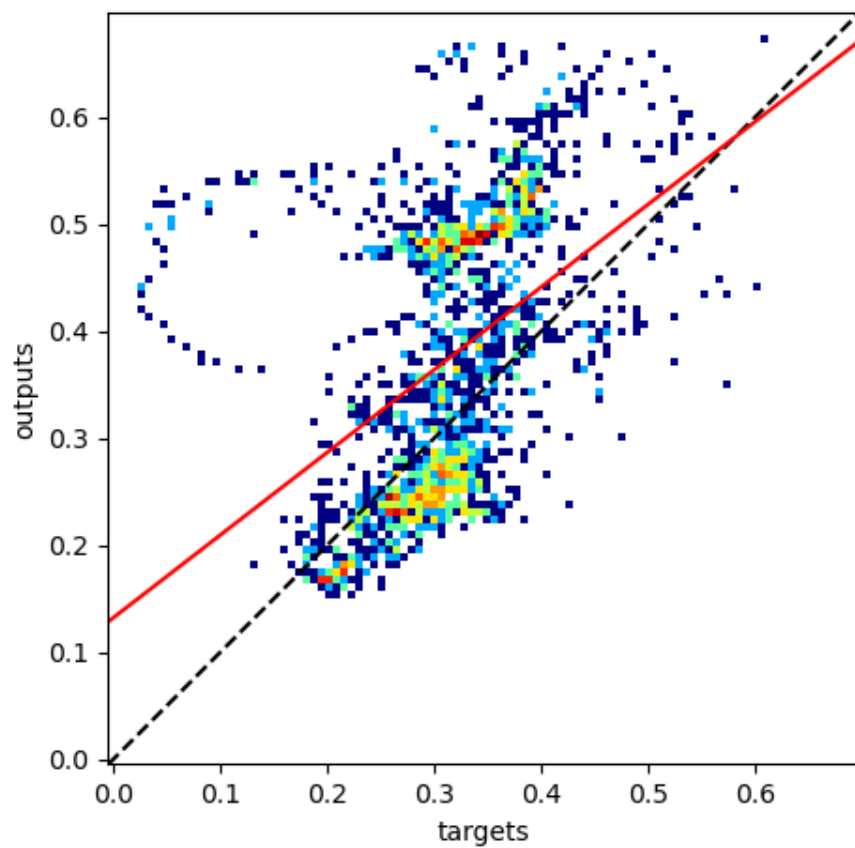
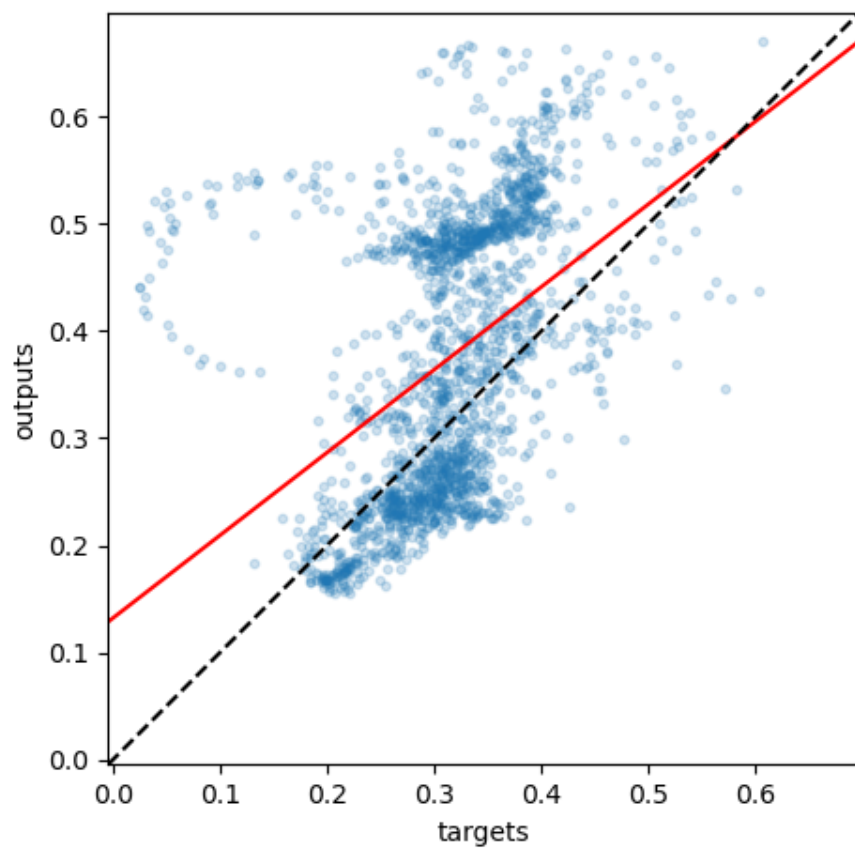


2021-04-10

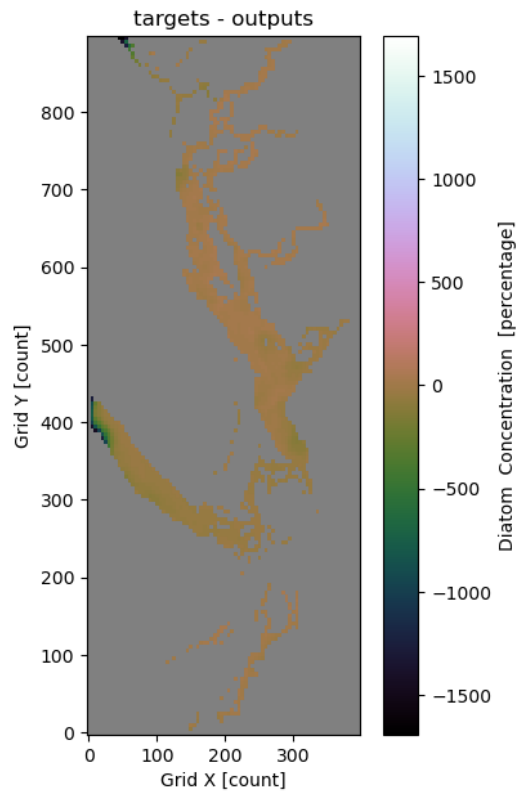
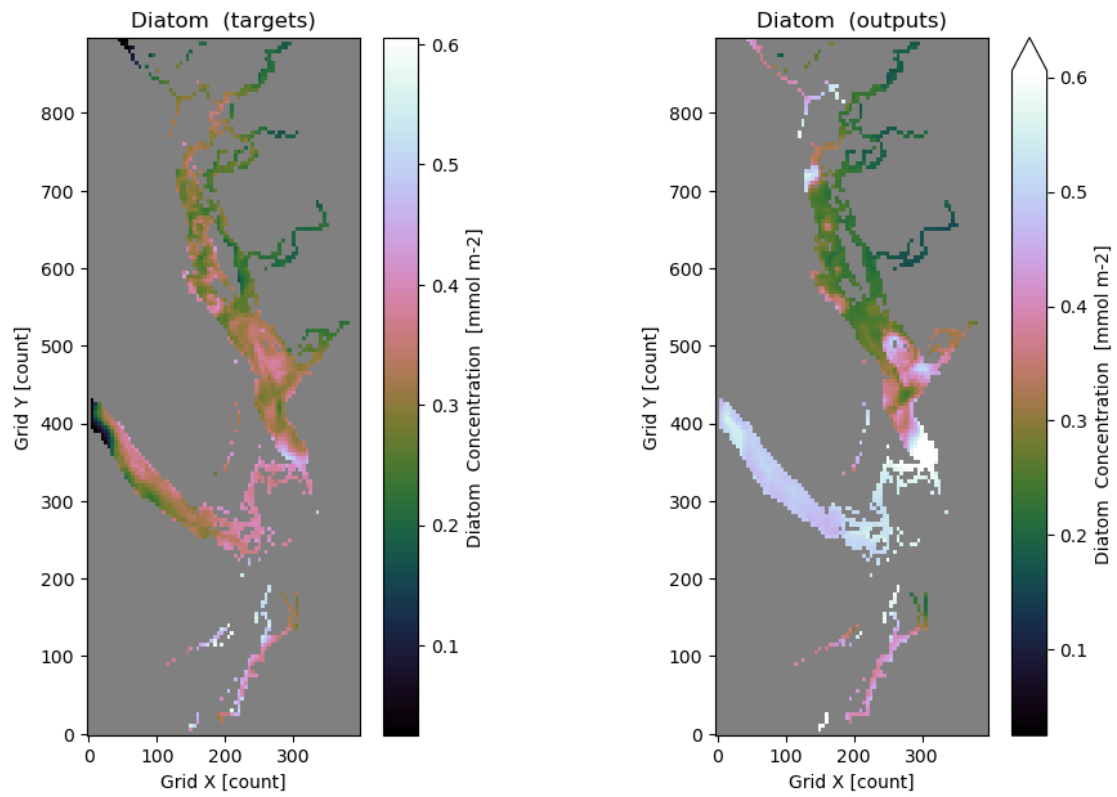


The amount of data points is 1863
The slope of the best fitting line is 0.773
The correlation coefficient is: 0.45
The mean square error is: 0.0174

Diatom 2021-03-13

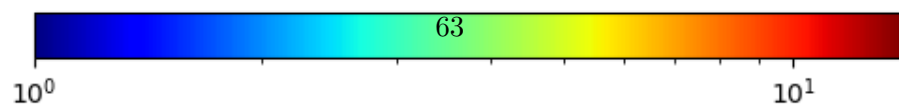
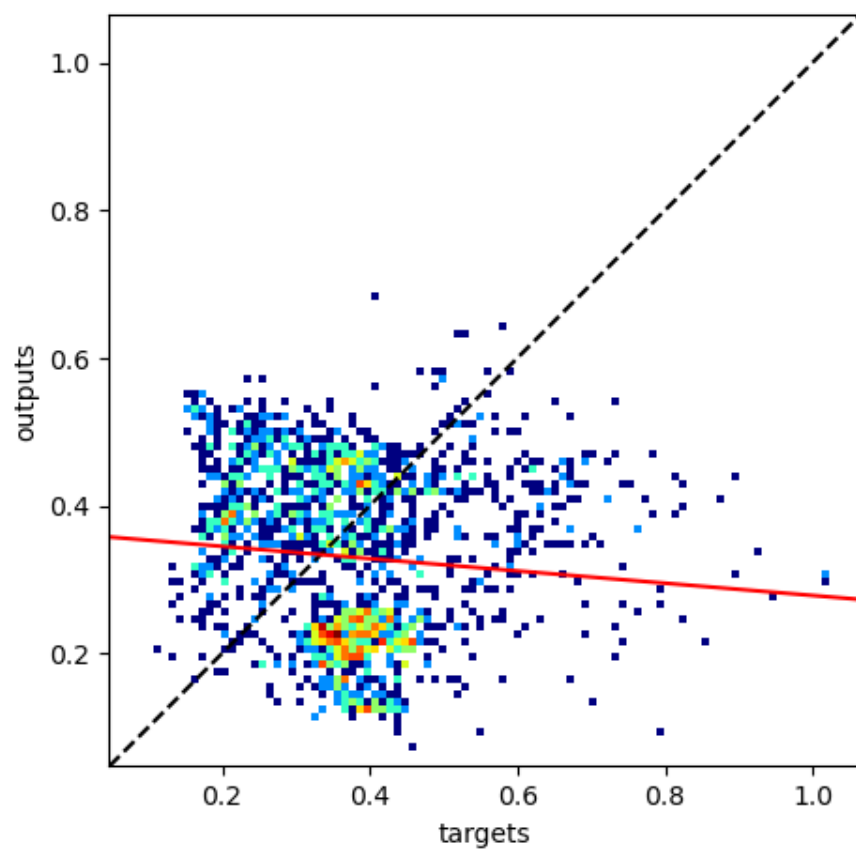
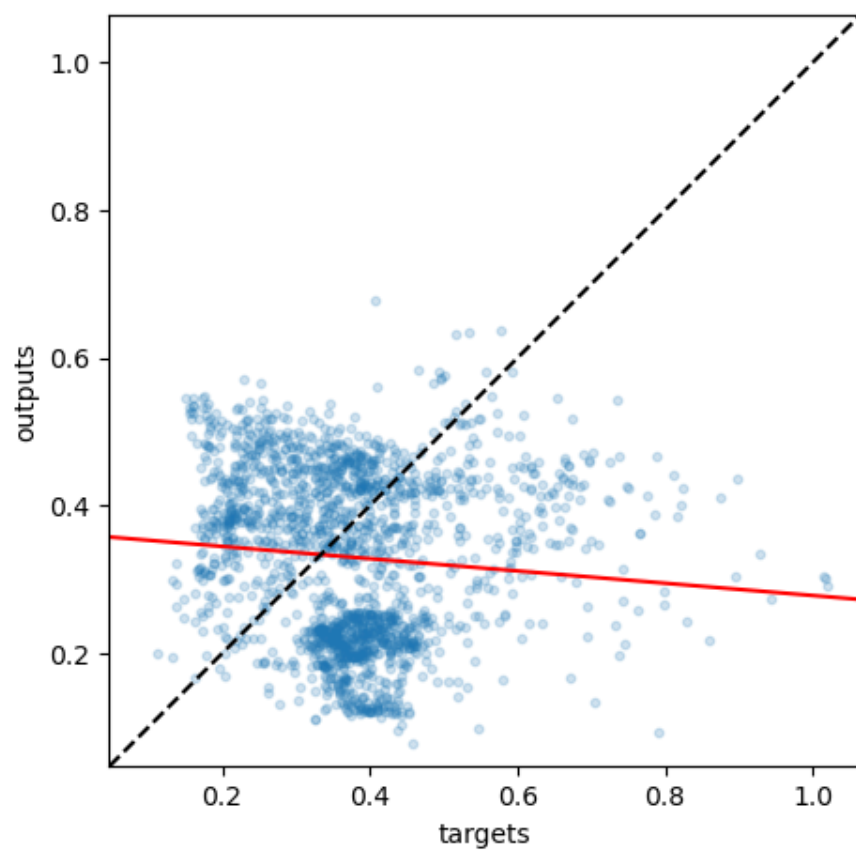


2021-03-13

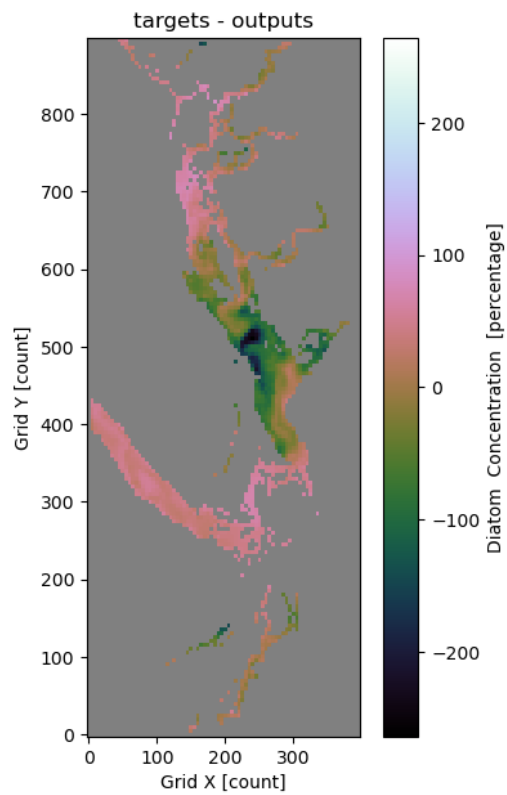
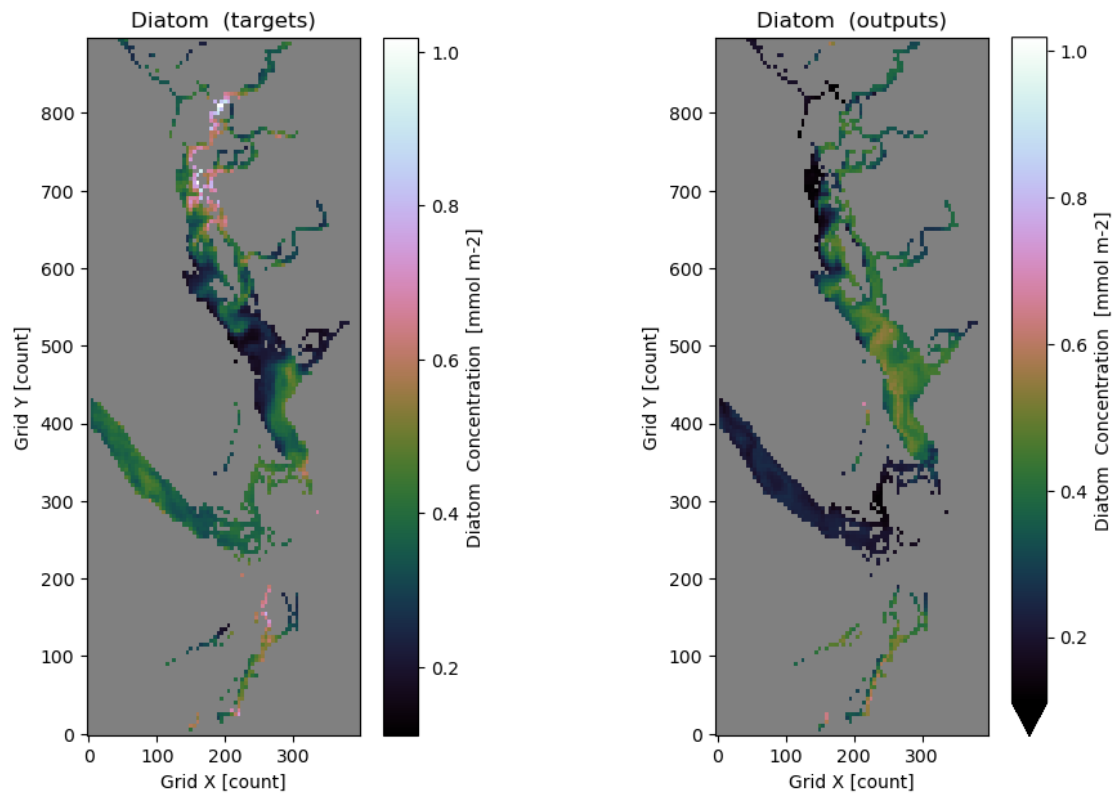


The amount of data points is 1863
The slope of the best fitting line is -0.083
The correlation coefficient is: -0.09
The mean square error is: 0.03353

Diatom 2008-04-28

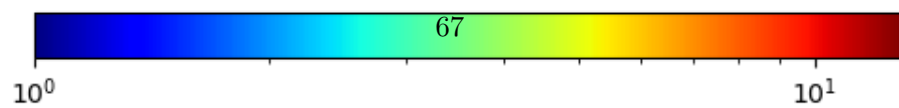
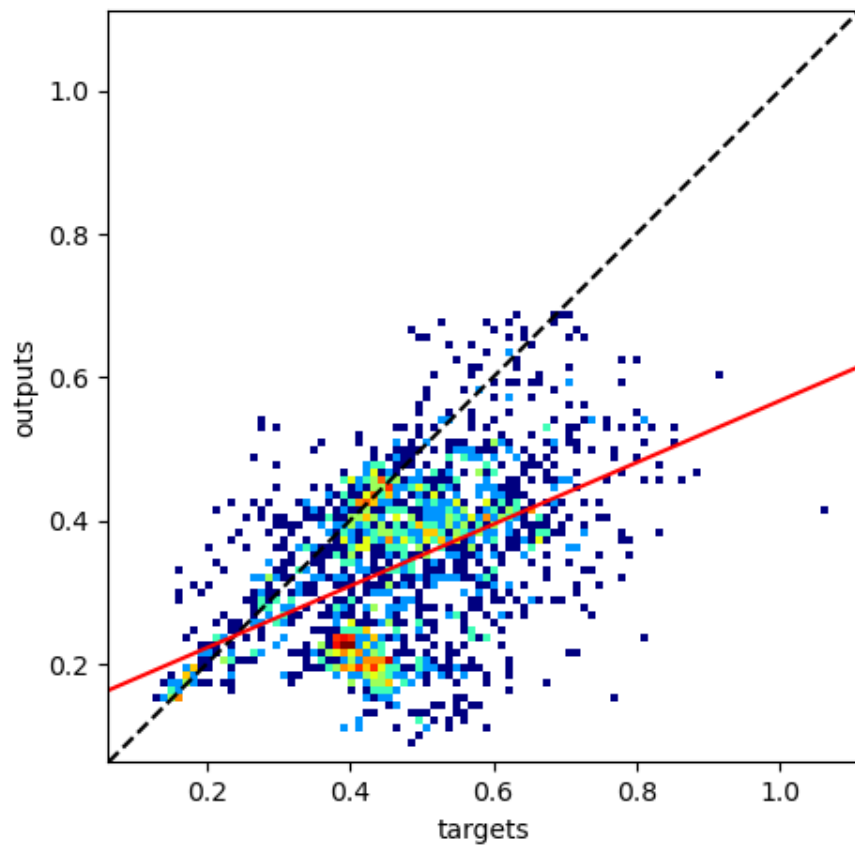
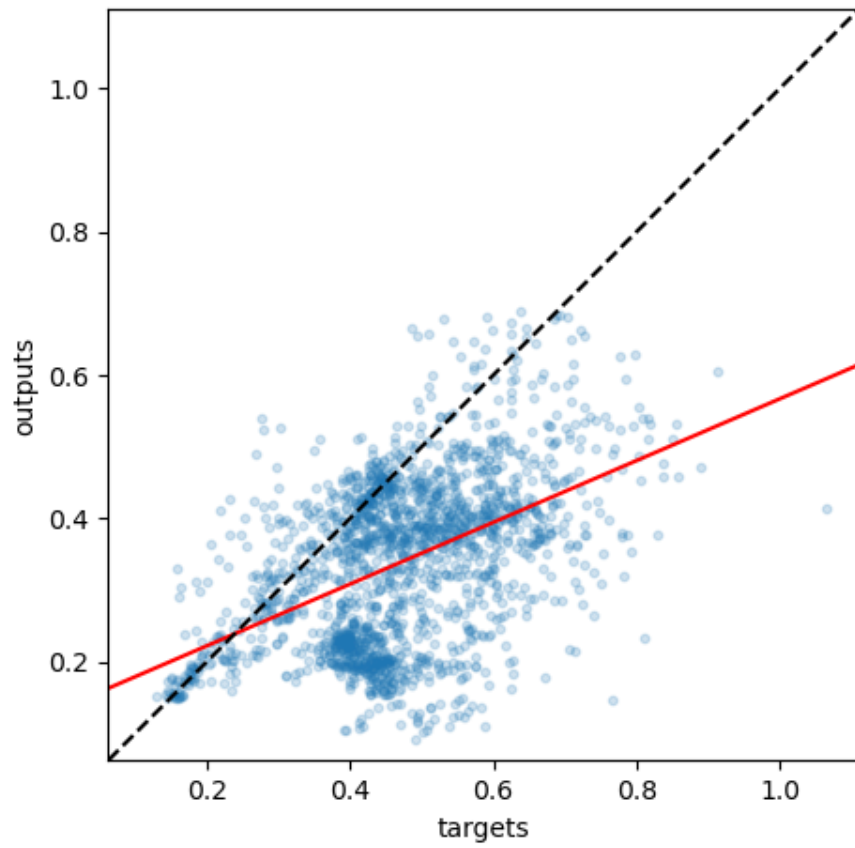


2008-04-28

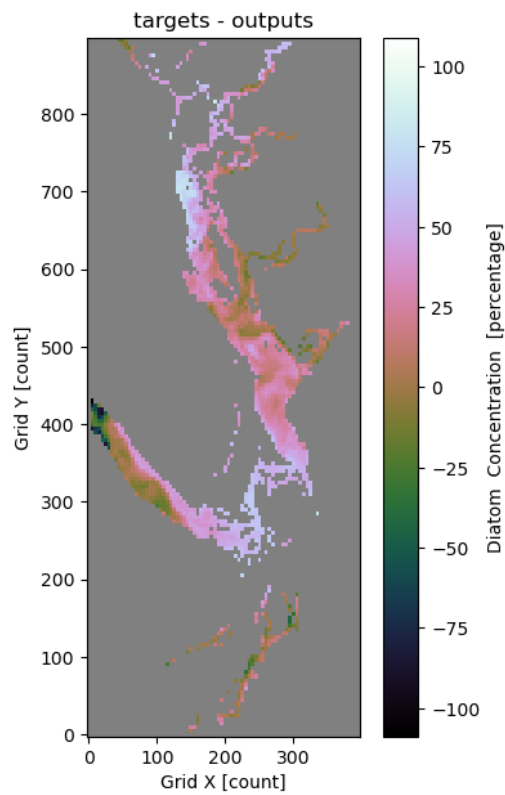
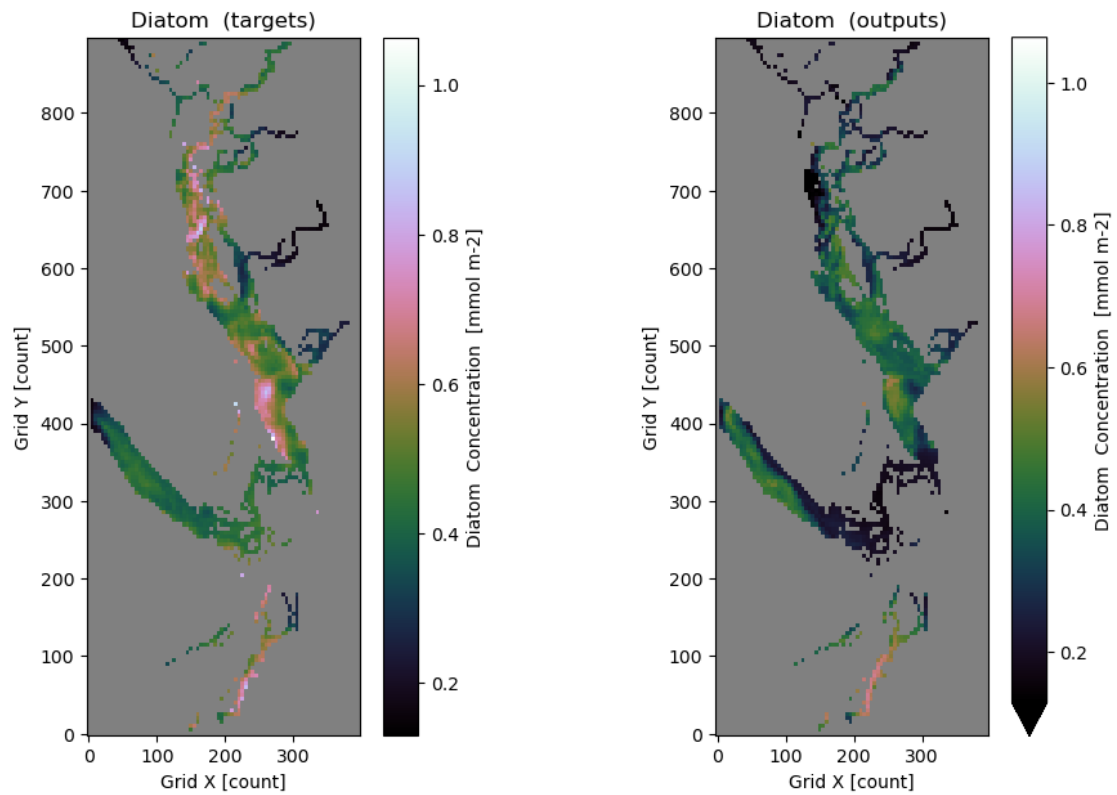


The amount of data points is 1863
The slope of the best fitting line is 0.431
The correlation coefficient is: 0.479
The mean square error is: 0.0332

Diatom 2014-04-03

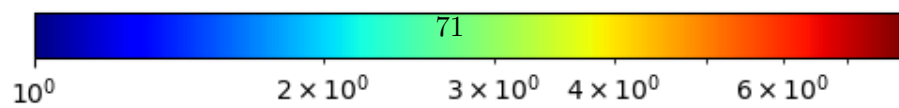
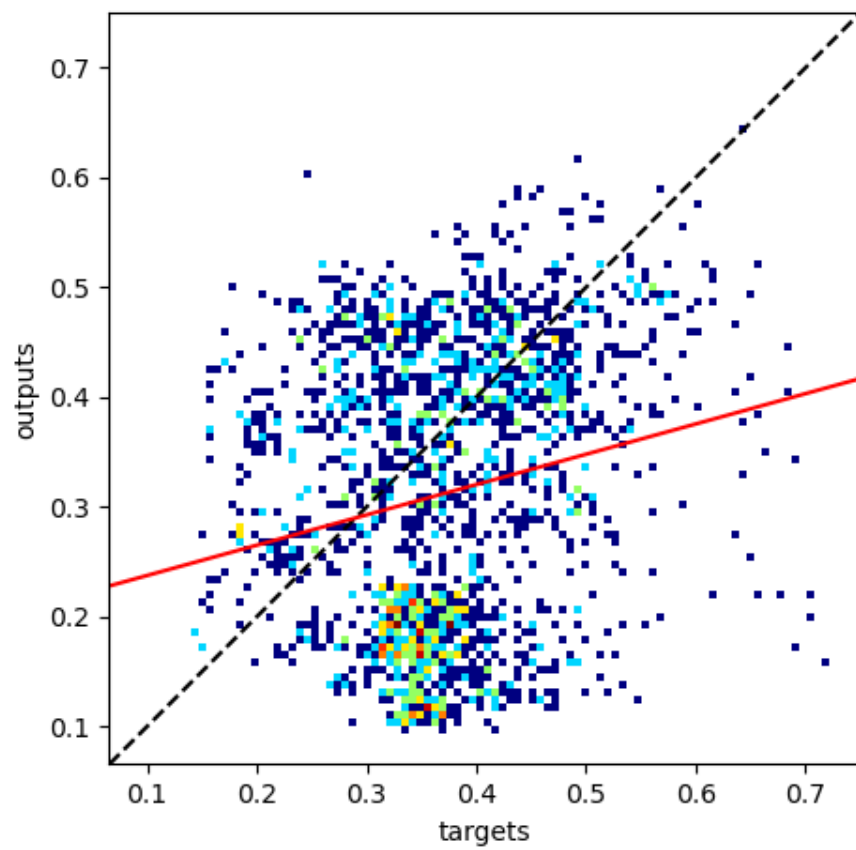
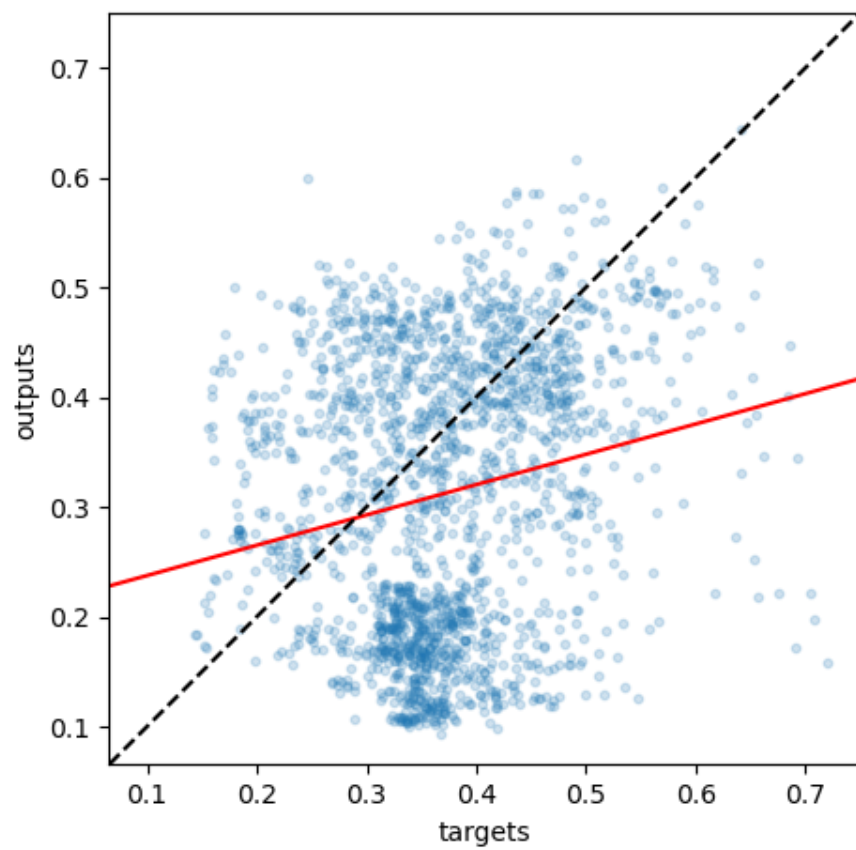


2014-04-03

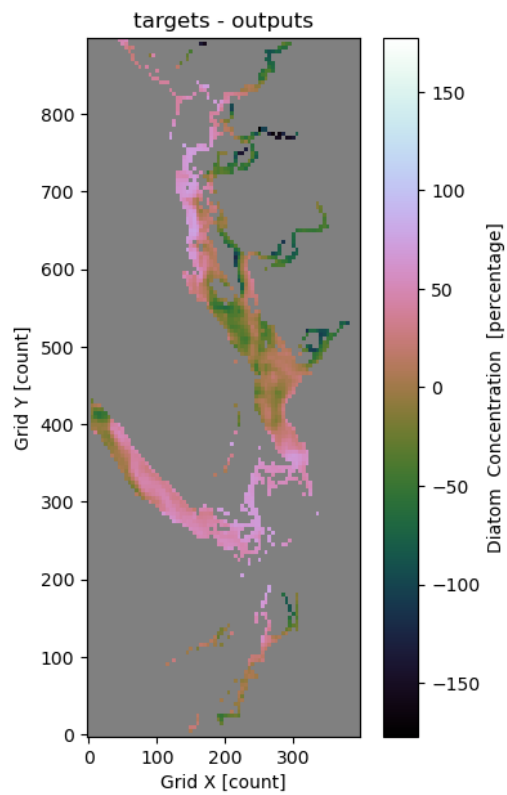
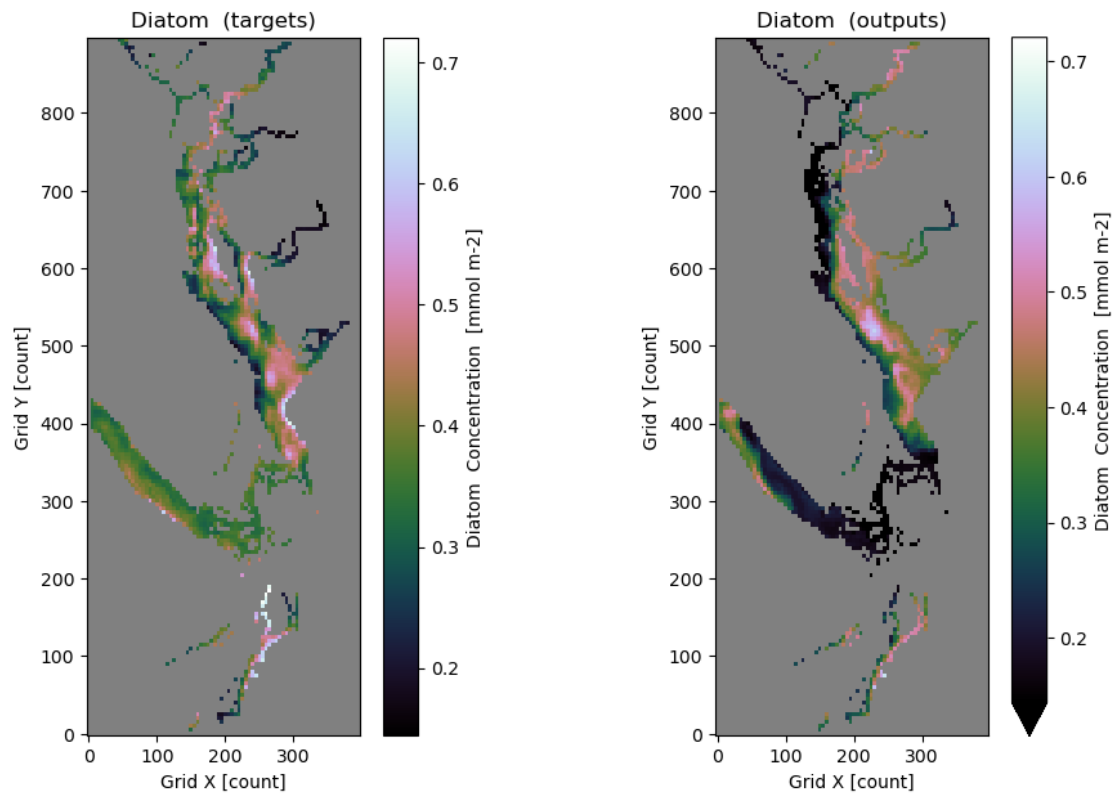


The amount of data points is 1863
The slope of the best fitting line is 0.276
The correlation coefficient is: 0.199
The mean square error is: 0.02332

Diatom 2014-04-27

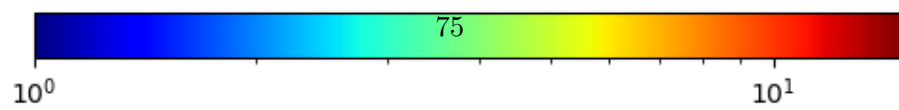
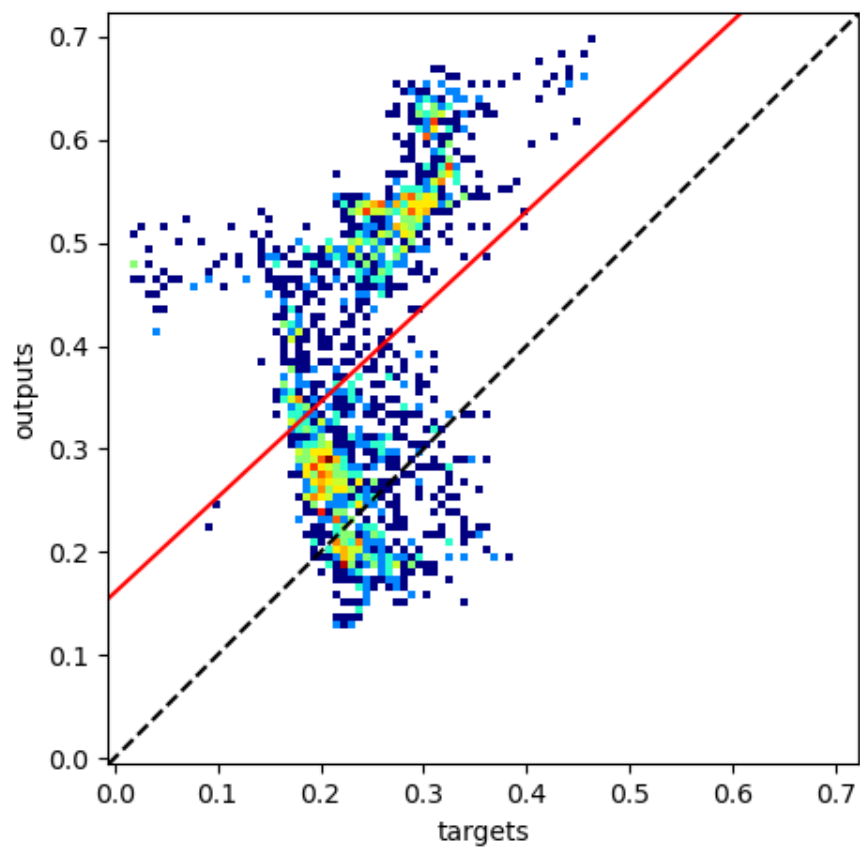
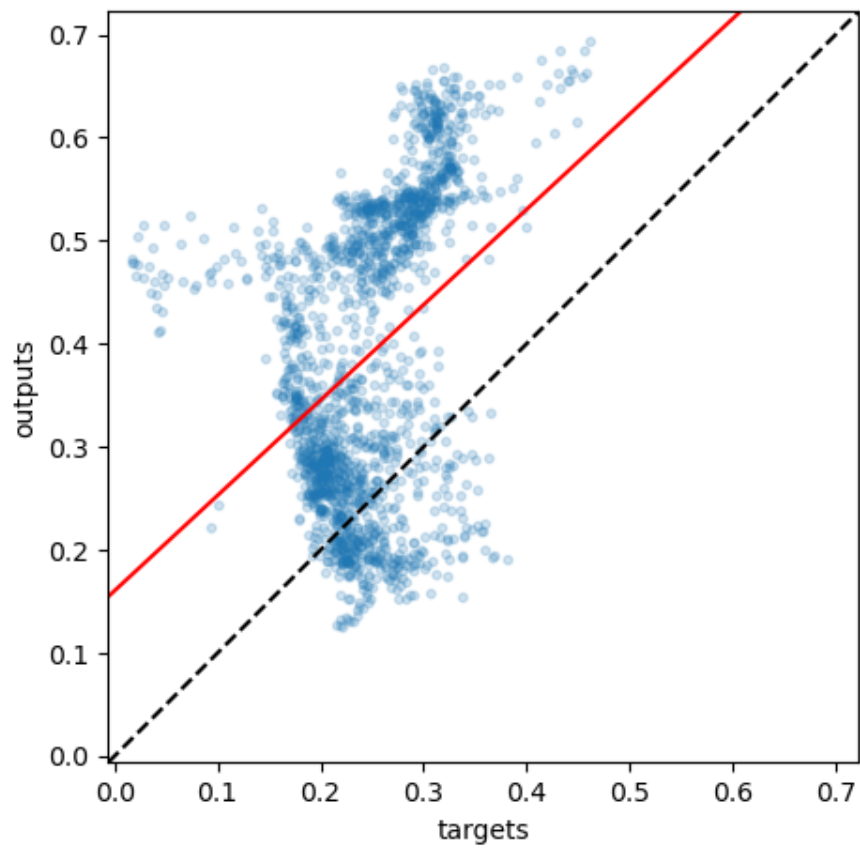


2014-04-27

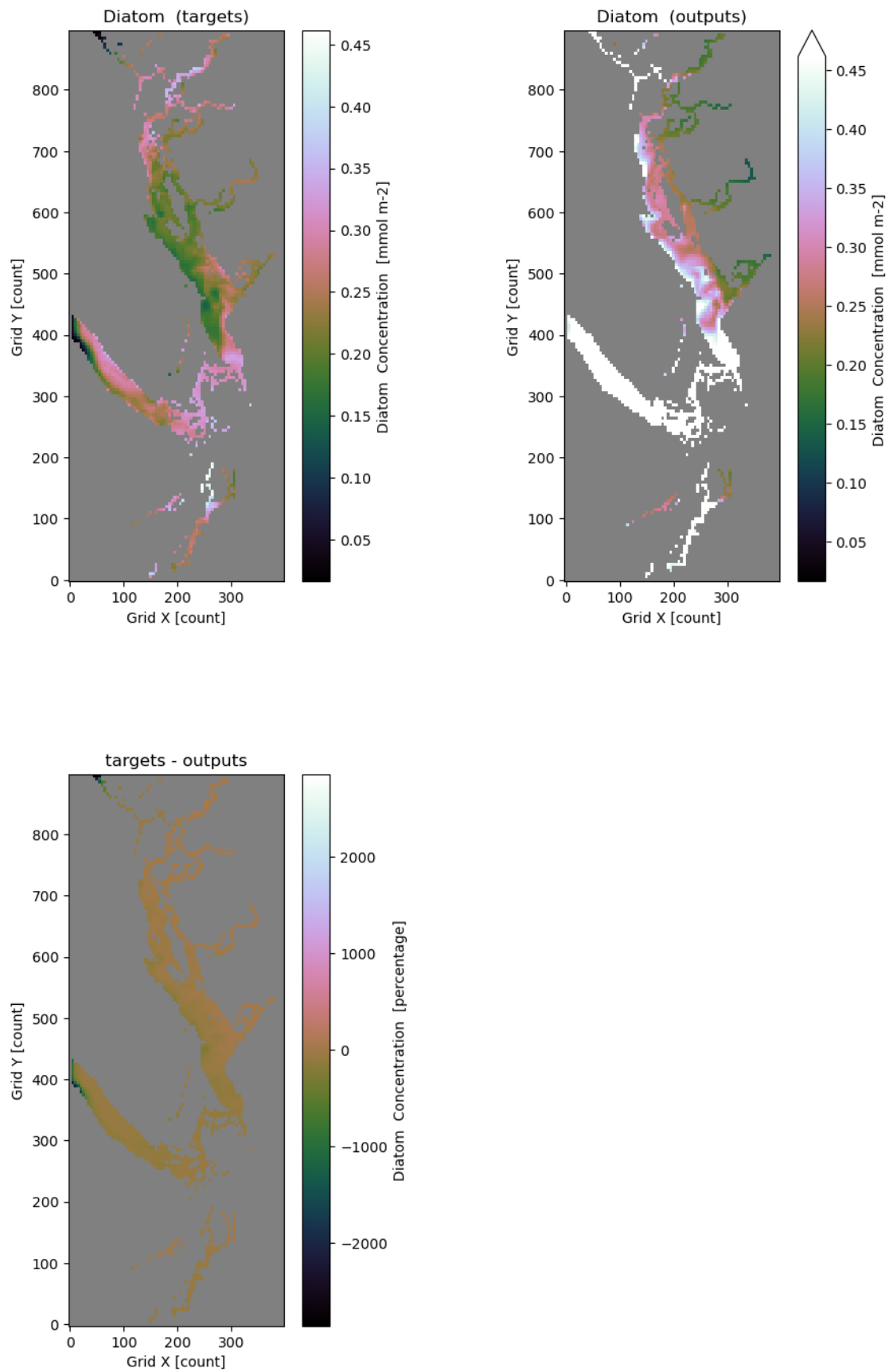


The amount of data points is 1863
The slope of the best fitting line is 0.925
The correlation coefficient is: 0.379
The mean square error is: 0.03772

Diatom 2018-03-03

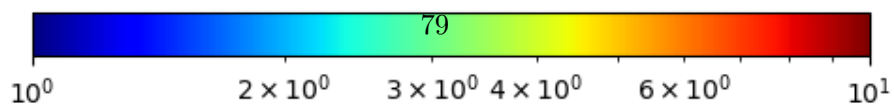
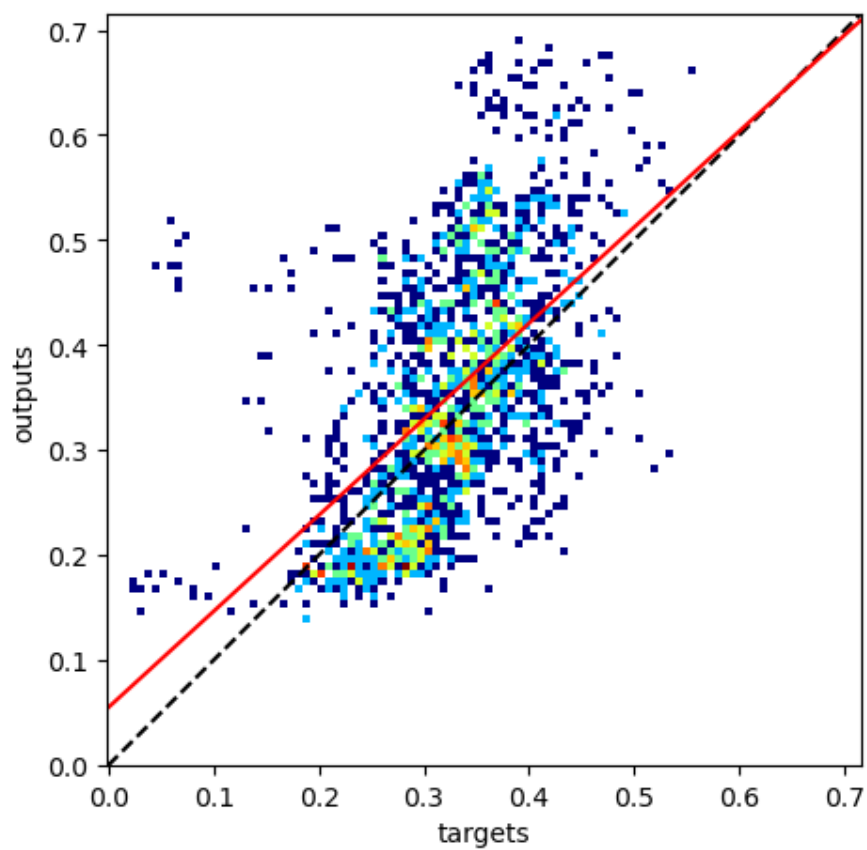
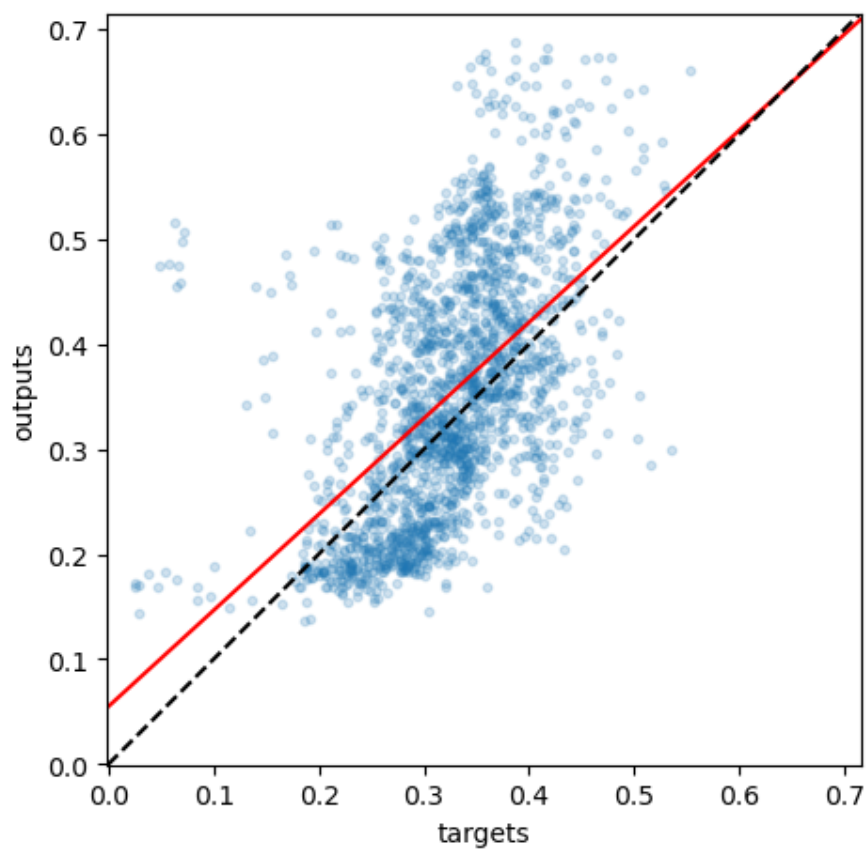


2018-03-03

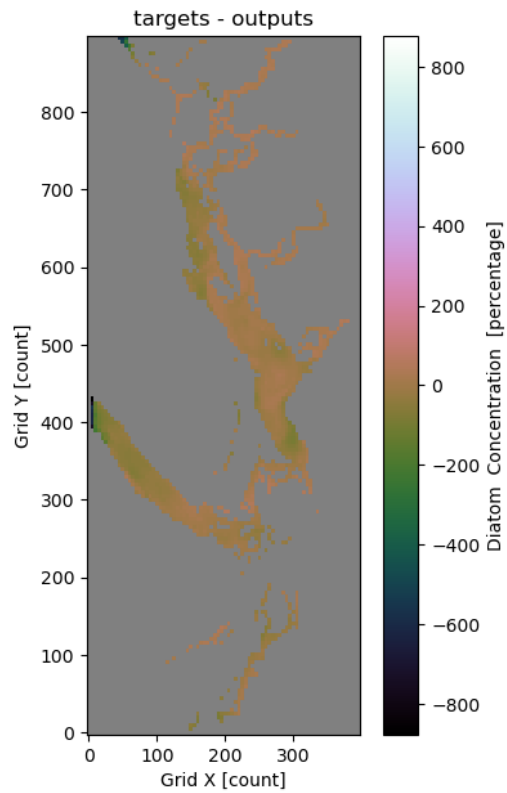
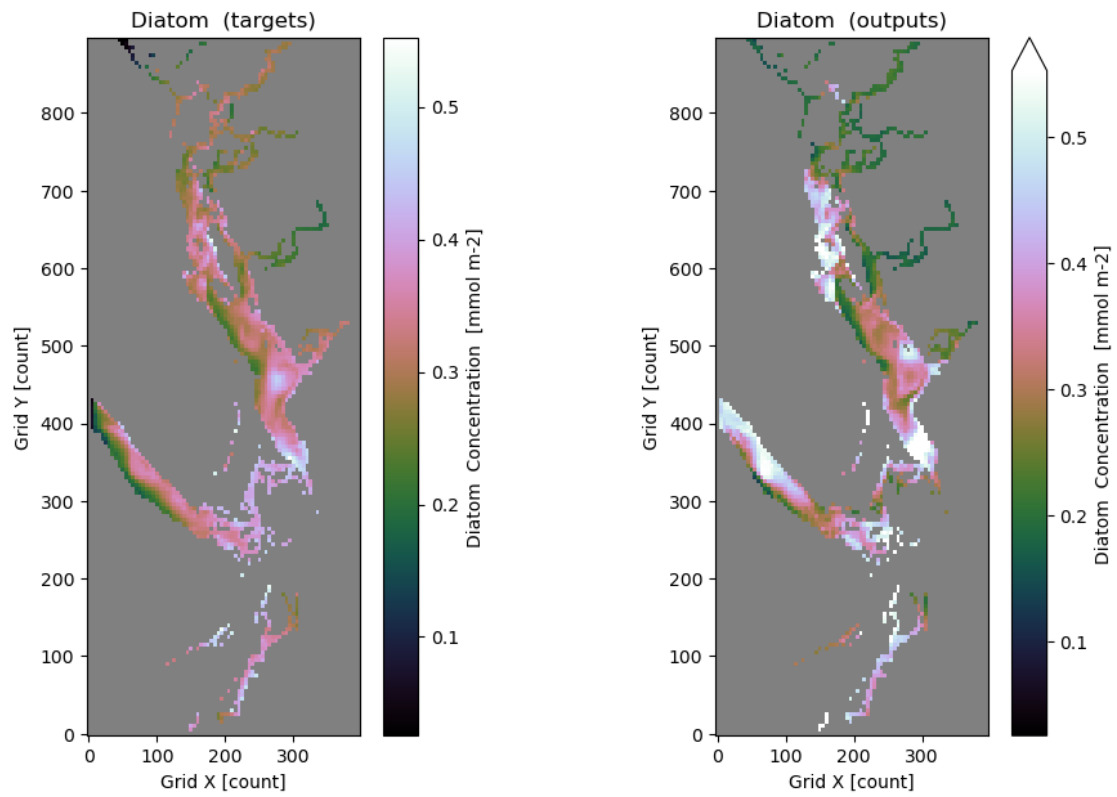


The amount of data points is 1863
The slope of the best fitting line is 0.915
The correlation coefficient is: 0.545
The mean square error is: 0.01061

Diatom 2014-03-14

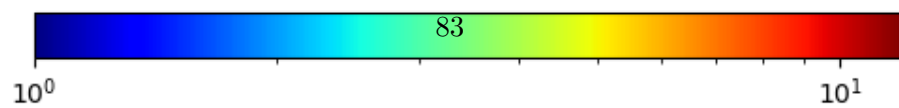
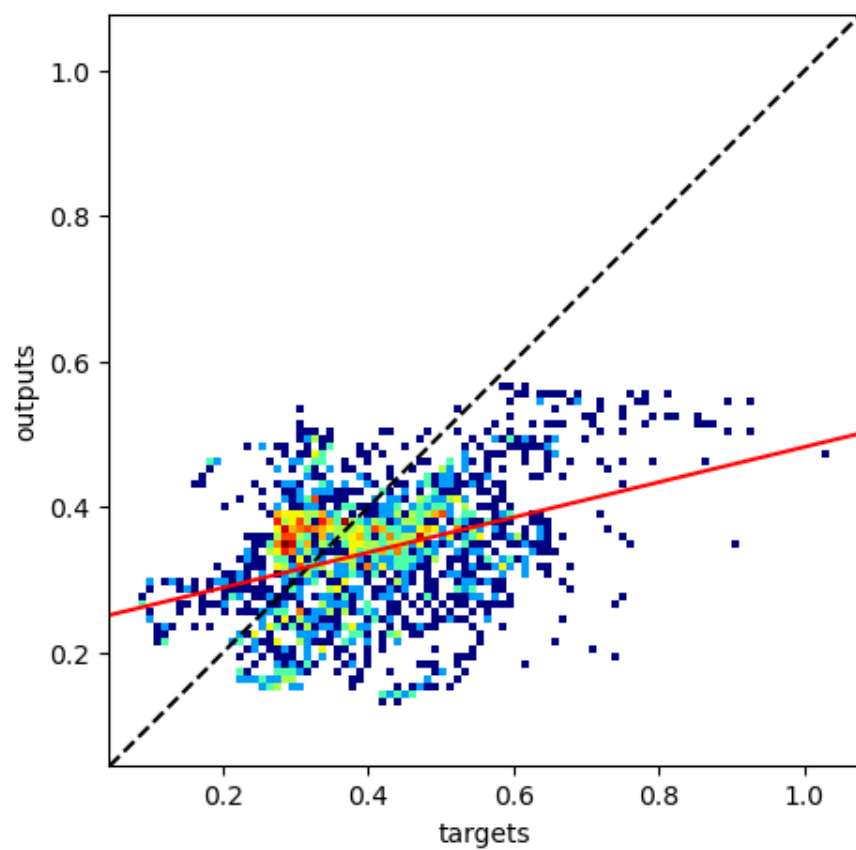
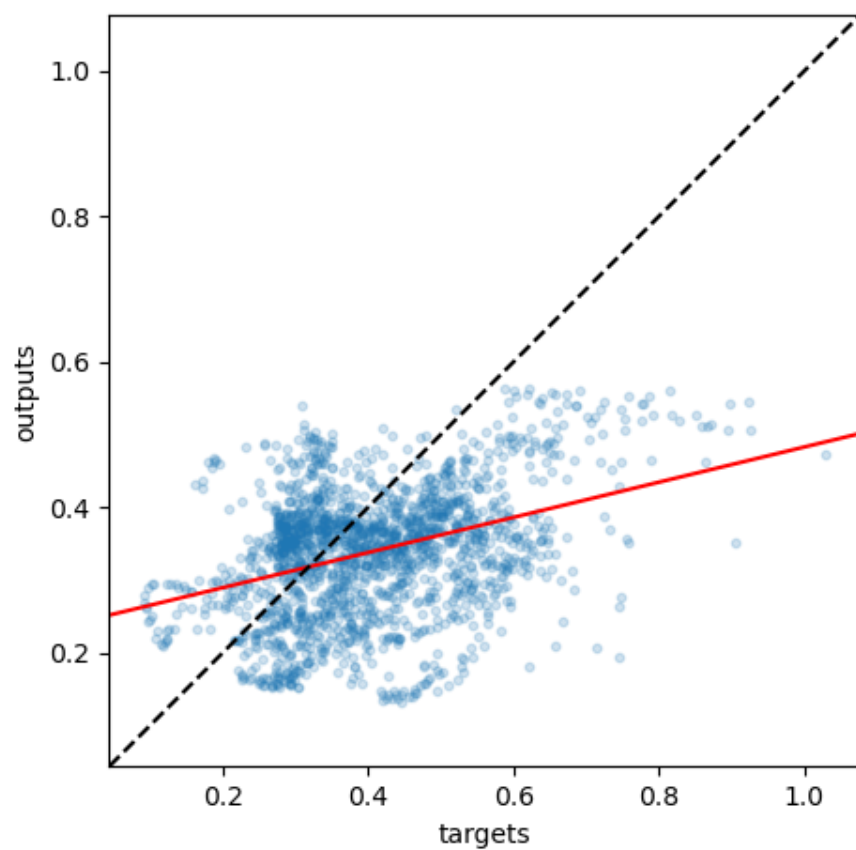


2014-03-14

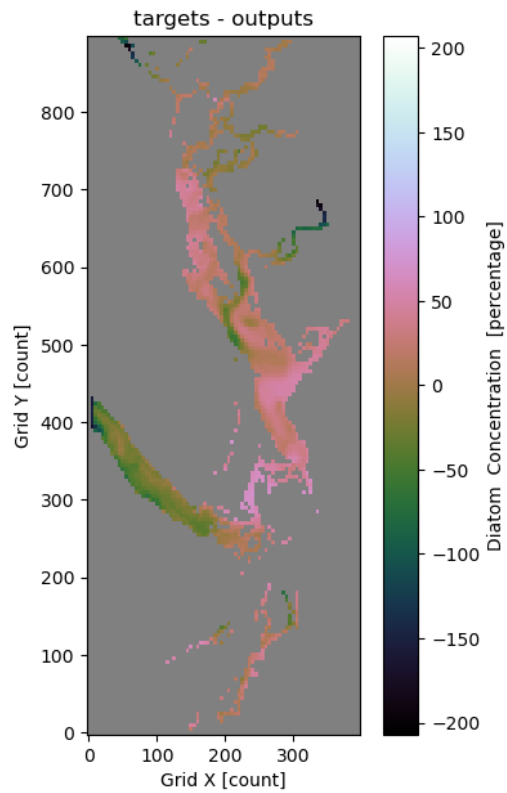
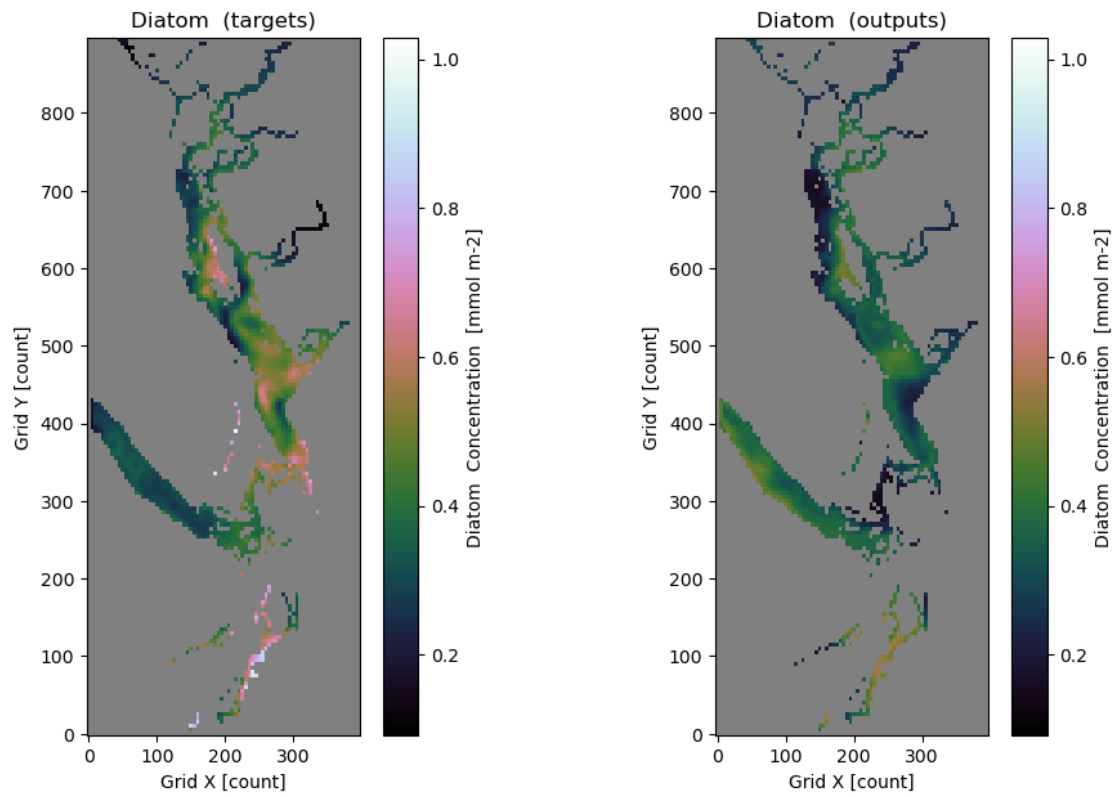


The amount of data points is 1863
The slope of the best fitting line is 0.242
The correlation coefficient is: 0.367
The mean square error is: 0.02019

Diatom 2015-04-04



2015-04-04



[]: