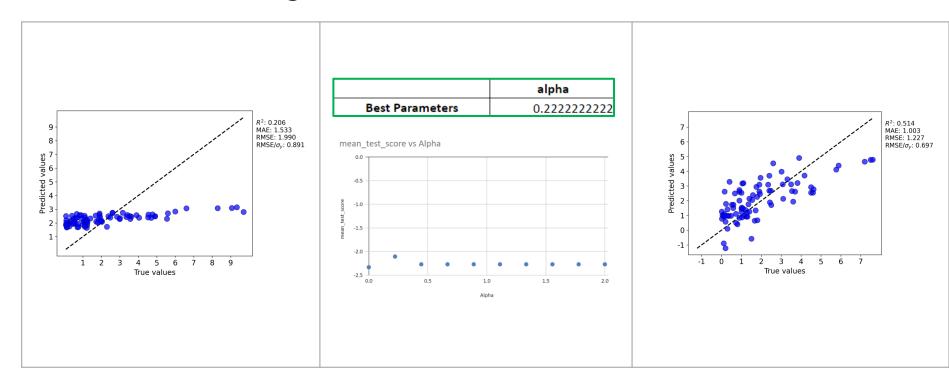
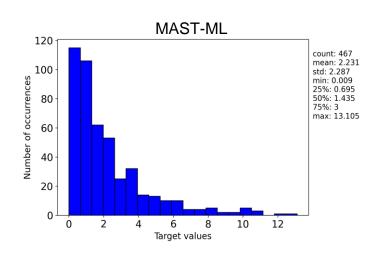
Assessment Figures

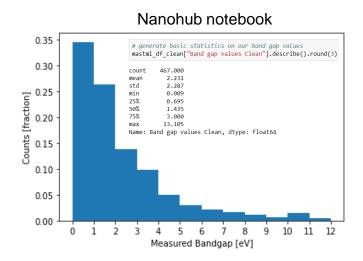


ML4ER - Assignment 5 Activities

Muhammad Zain Azeem, Informatics Skunkworks (**non-credits**), Week 3 04/08/2024

Let's look through the outputs and compare them to some of the initial dataset analysis and compare to the previous Nanohub workflow. Open Question: the "histogram_target_values.png" file in the newly created DataCleaning folder under our output directory. Compare back to the histogram we made in the previous notebook. Are they the same?





Conclusion: Results were found to be similar

Question:

Using the cell block below with outputs the feature object directly compare the features generated to those in the previous workflow. Do we have the same total number?

If they're different can you think of any reasons why?

hint: mastml does some initial cleaning automatically on the features.

MAST-ML

	AtomicNumber_composition_average	AtomicRadii_composition_average
	41.000000	1.292000
1	32.000000	1.217000
2	17.000000	1.051000
3	49.500000	1.522000
4	34.000000	1.206000
462	36.000000	1.466667
463	35.500000	1.450000
464	32.000000	1.382500
465	35.428571	1.475714
466	23.000000	1.255000
467 rd	ows × 87 columns	

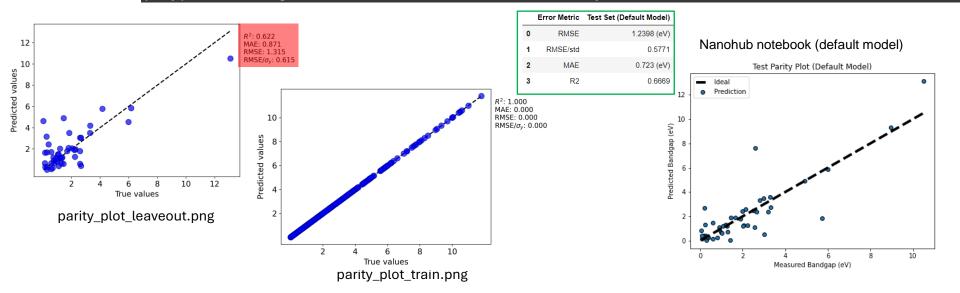
Nanohub notebook

	AtomicNumber_composition_average	AtomicRadii_composition_average
0	6.000000	1.135000
1	10.000000	1.270000
2	19.000000	1.345000
3	15.000000	1.560000
4	28.000000	1.440000
462	60.500000	1.422500
463	83.000000	1.700000
464	35.333333	1.086000
465	50.000000	1.057500
466	36.000000	0.948333

467 rows × 87 columns

Conclusion: Different results were obtained

After this run completes we want to go look at how the model is performing. Navigate to the newly created "DecisionTreeRegressor..." folder Question: and find both the "parity_plot_leaveout.png" file as well as the "parity_plot_train.png" file. Compare them both to eachother as well as to the parity plots made during the Nanohub notebook for the default model. Are they the same? Similar?



Conclusion: Nanohub results showed higher R²: 0.669 then MAST-ML R²: 0.622

Task:

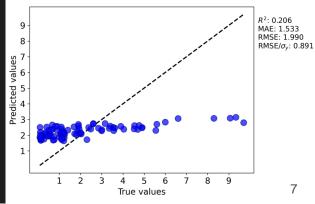
- 1) pick another model type from scikit-learn. You can see a reference for available models here: https://scikit-learn.org/stable/supervised_learning.html If you're not sure what kind of model to try I might suggest one of the linear type models such as Ridge Regression or LASSO. To see the list of available hyperparameters for each model you can click their respective link.
- 2) build a default model where you don't change any hyperparameters from the scikit-learn defaults and analyze it's performance both on the Test data and with a 5-fold CV

```
# Selection of model "LASSO"
default LASSO = SklearnModel(model='Lasso')
models = [default LASSO]
selector = [NoSelect()]
metrics = ['r2_score', 'mean_absolute_error', 'root_mean_squared_error', 'rmse_over_stdev']
processor = SklearnPreprocessor(preprocessor='StandardScaler', as frame=True)
# Set up the 5-fold cross-validation with 2 repeats
splitter = SklearnDataSplitter(splitter='RepeatedKFold', n_repeats=2, n_splits=5)
# Evaluate the model using cross-validation and test set
splitter.evaluate(X=X,
                  models=models.
                  preprocessor=processor,
                  selectors=selector,
                  metrics=metrics,
                  savepath=savepath,
                  X extra=X extra,
                  leaveout inds=X testdata,
```

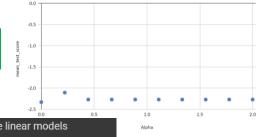
verbosity=3)

- 1) Selected "LASSO" model
- 2) Performed 5-CV and splitting

Results:



	alpha
Best Parameters	0.222222222



Task:

3) perform a grid search on 1 of the hyperparameters. I'd suggest picking the alpha hyperparameters if using one of the linear models suggested above.

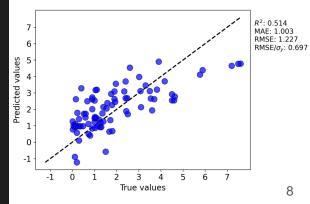
```
# Activity 5
# Selection of model "LASSO" - Optimization
default_LASSO = SklearnModel(model='Lasso',alpha=0.22)
models = [default LASSO]
selector = [NoSelect()]
metrics = ['r2 score', 'mean absolute error', 'root mean squared error', 'rmse over stdev']
processor = SklearnPreprocessor(preprocessor='StandardScaler', as frame=True)
# Set up the 5-fold cross-validation with 2 repeats
splitter = SklearnDataSplitter(splitter='RepeatedKFold', n_repeats=2, n_splits=5)
# Evaluate the model using cross-validation and test set
splitter.evaluate(X=X,
                  y=y,
                  models=models,
                  preprocessor=processor,
                  selectors=selector,
                  metrics=metrics,
                  savepath=savepath,
                 X extra=X extra,
                  leaveout inds=X testdata,
                  verbosity=3)
```

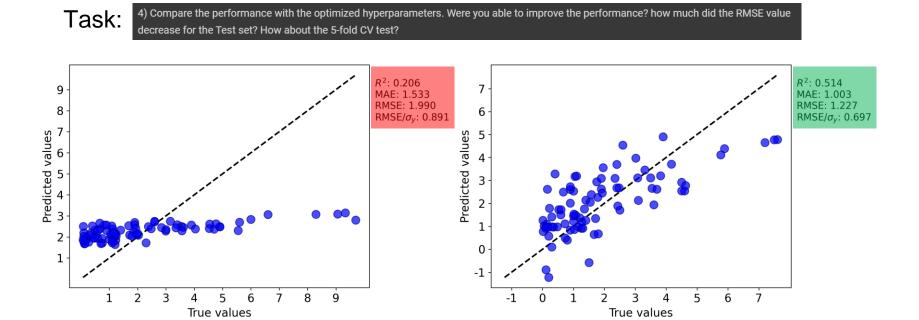
3) Performed grid search

mean_test_score vs Alpha

4) Performed 5-CV and splitting

Results:





Conclusion: The model performed improved from R²: 0.206 to R²: 0.514