

Integration of Meta Machine Learning in GenoML

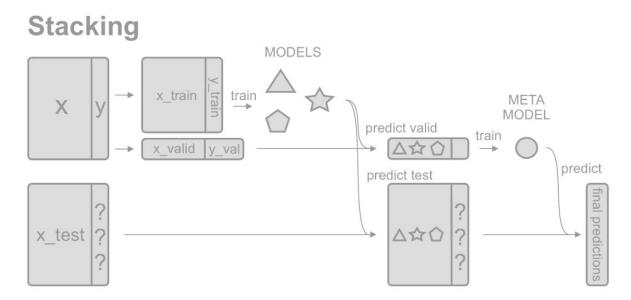
Authors:

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- 1. Meta Machine Learning.
- 2. Code.
- 3. Using the tool.

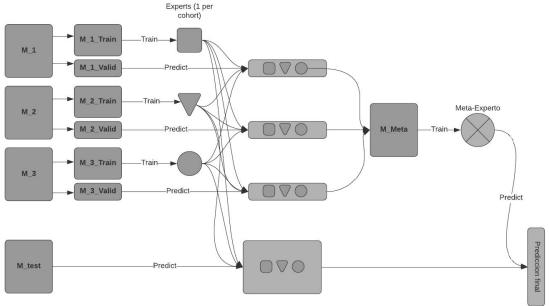
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Meta-ml as a basic algorithm in GenoML



Meta-ML as a multi-silo integration algorithm in

GenoML



Loss of accuracy for using a multi-silo approach with nSmE

Algorithm	Mean	Confidence interval
QuadraticDiscriminantAnalysis	50.3%	48.3 - 52.3
RandomForestClassifier	56.7%	54.5 - 58.9
MML nSmE	62.5%	58.7 - 66.2
LogisticRegression	71.7%	71.7 - 71.7
SGDClassifier	72.1%	68.5 - 75.7

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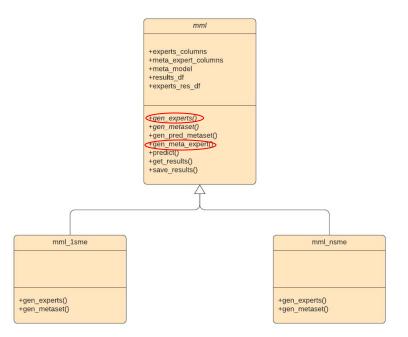
Code

Class hierarchy:

- Abstract class mml
- Classes inheriting from mml

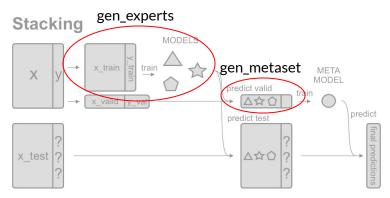
gen_experts and gen_meta_expert use GenoML base functionality.

https://github.com/edusalcas/TFG MML GenoML



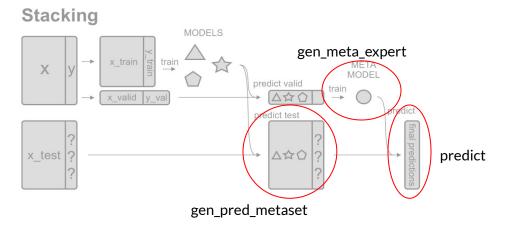
Code - mml abstract methods

```
68
69
          @abstractmethod
70
          def gen_experts(self,
                          path,
                          expert_names=None,
                          prefix=None,
                          metric_max=None,
75
                          seed=None,
76
                          metric tune=None,
                          max_tune=None,
                          algs=None):
78
79
80
81
          @abstractmethod
82
          def gen_metaset(self,
83
                          data,
84
                          experts_path,
                          experts names):
86
87
```



Code - mml methods

```
def gen_pred_metaset(self, data, experts_path, experts_names, path, meta_prefix): def gen_meta_expert(self, data, path, experts_names, prefix, metric_max, algs, seed, metric_tune, max_tune): def predict(self, data, path, experts_names, meta_prefix, iteration=0): def predict(self, data, path, experts_names, path, experts_names, path, experts_names, path, experts_names, path, experts_names
```



Code - mml_1sme and mml_nsme methods

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Workflow

- 1. Munge data.
- 2. Split train-test.
- 3. Generate level one experts.
- 4. Generate meta-expert.
- 5. Predict.

- 1. Munging from GenoML.
- 2. train_test_split from sklear
- 3. gen_experts from our code
- 4. gen_meta_expert from our code
- 5. predict from our code

In points 3 and 4, we use train and tune functions from GenoML.

Example - Experiment with MML 1SmE

```
manager_mml = MML 1SmE(actual_path + prefix)
manager_mml.gen_experts(path=actual_path,
                        expert names=algorithms,*
                        prefix=train_prefix,:
                        metric max=metric,*
                        seed=seed,
                        metric_tune=metric,*
                        max tune=tune iterations)
manager_mml.gen_meta_expert(data=dataForML_test,
                            path=actual_path,
                            experts_names=algorithms,
                            prefix=meta prefix,>
                            metric_max=metric,
                            algs=algorithms_names,*
                            seed=seed,
                            metric_tune=metric,*
                            max tune=tune iterations);
balacc = manager_mml.predict(data=test_df,
                             path=actual_path,
                             experts names=algorithms,
                             meta_prefix=meta_prefix,
                             iteration=n)
```

- *: Parameters needed for base GenoML functions
- *: Parameters needed for modified GenoML Train Function.



End

Thanks for your attention