## custom\_assessment\_function\_with\_search\_grid

## February 17, 2019

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In [1]: from sklearn.metrics import log_loss, make_scorer
        Log_loss = make_scorer(log_loss, greater_is_better=False, needs_proba=True) # logarith
In [2]: from sklearn.datasets import load_digits
        digits = load_digits()
        X, Y = digits.data, digits.target
        from sklearn import svm
        hp = svm.SVC(probability=True, random_state=1)
        from sklearn.model_selection import GridSearchCV
        search_grid = [
            {'C': [1, 10, 100, 1000], 'kernel':['linear']},
            {'C': [1, 10, 100, 1000], 'gamma': [0.001, 0.0001], 'kernel':['rbf'] },
        1
In [3]: search_func = GridSearchCV(estimator=hp, param_grid=search_grid, scoring=Log_loss, n_je
        search_func.fit(X, Y)
        print(search_func.best_score_)
        print(search_func.best_params_)
-0.16138394081976534
{'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
In [4]: # new lose function
        import numpy as np
        from sklearn.preprocessing import LabelBinarizer
        def custom_log_loss_func(ground_truth, p_predictions, penalty=list(), eps=1e-15):
            adj_p= np.clip(p_predictions, eps, 1-eps)
            lb = LabelBinarizer()
            g = lb.fit_transform(ground_truth)
            if g.shape[1] == 1:
                g.append(1-g, g, axis=1)
            if penalty:
                g[:,penalty] = g[:,penalty] * 2
            summation = np.sum(g * np.log(adj_p))
            return summation * (-1.0/len(ground_truth))
```

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In [5]: # scorrer thats add penalty for 4 and 9 becouse thoes numbers are easy to confused wit
        my_custom_scorer = make_scorer(custom_log_loss_func, greater_is_better=False, needs_pre-
In [7]: search_grid = [
            {'C': [1, 10, 100, 1000], 'kernel':['linear']},
            {'C': [1, 10, 100, 1000], 'gamma': [0.001, 0.0001], 'kernel':['rbf'] },
        ]
        search_func = GridSearchCV(estimator=hp, param_grid=search_grid,
                                   scoring=my_custom_scorer,
                                   n_jobs=1, iid=False,
                                   refit=True, cv=3)
        search_func.fit(X, Y)
        print(search_func.best_score_)
        print(search_func.best_params_)
-0.19961027129825612
{'C': 1, 'gamma': 0.001, 'kernel': 'rbf'}
In []:
```