

```
In [1]: import pandas as pd
import os
import numpy as np
from sklearn.model_selection import RepeatedStratifiedFold
from fastai import *
from fastai.vision import *
from fastai.callbacks import *
```

For loop Repeated Stratified K-Fold

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In [2]: df_Test_Train = pd.read_csv('Fastai_dataset_usable.csv')
df_Test_Train.drop(df_Test_Train.columns[df_Test_Train.columns.str.contains('unnamed',case = False)],axis = 1, inplace = True)
df_Test_Train.head(100)
```

Out[2]:

	Crop_Filepath	clutch	Day	egg_number	sex
0	Cropped_Egg_images/Clutch1_D18egg2.JPG	1	18	2	Female
1	Cropped_Egg_images/Clutch1_D18egg3.JPG	1	18	3	Female
2	Cropped_Egg_images/Clutch1_D18egg4.JPG	1	18	4	Male
3	Cropped_Egg_images/Clutch1_D18egg6.JPG	1	18	6	Female
4	Cropped_Egg_images/Clutch1_D18egg9.JPG	1	18	9	Female
5	Cropped_Egg_images/Clutch1_D18egg11.JPG	1	18	11	Female
6	Cropped_Egg_images/Clutch1_D18egg15.JPG	1	18	15	Female
7	Cropped_Egg_images/Clutch1_D18egg16.JPG	1	18	16	Male
8	Cropped_Egg_images/Clutch1_D18egg17.JPG	1	18	17	Female
9	Cropped_Egg_images/Clutch1_D18egg19.JPG	1	18	19	Male
10	Cropped_Egg_images/Clutch1_D18egg20.JPG	1	18	20	Male
11	Cropped_Egg_images/Clutch1_D18egg21.JPG	1	18	21	Female

```
In [3]: ### For Loop
'''Things we want:
-arches: Resnet34, Resnet34
-wds: .01, .003, 1
-transforms: None, Modified
-normalization: True and False
-folds:5
-repeats:5
'''

# Parameters to vary
model_archs = [models.resnet34, models.resnet50]
weight_decay = [.01, .003, 1]
normalization = [True, False]
epoch_cycles=1
np.random.seed(42)
modified = get_transforms(do_flip = True, flip_vert = False, max_rotate = 35, max_lighting = None, max_warp = -.2, p_lighting = 0)

# Tests to Perform

tests = [[models.resnet34, .01, None, False],
[models.resnet34, .003, modified, False],
[models.resnet50, 1, modified, True],
[models.resnet50, .01, modified, True]]

# Creating Frame Work
DFBig = pd.DataFrame(columns = ['test_name', 'model_arch', 'transforms', 'normalized',
                               'weight_decay', 'split_num', 'max_error', 'min_error', 'avg_error', 'train_df'])

# Creating Stratified K folds
rskf = RepeatedStratifiedFold(n_splits=5, n_repeats=4)

# For Loop for test
split_num = 1
for s in tests:
    wd = s[1]
    norm = s[2]
    arch = s[3]
    archstr = str(arch).split(' ')[1]
    if s[2] is not None:
        tsfstr = 'modified'
    else:
        tsfstr = 'none'
    test_name = archstr + '_' + str(wd) + '_' + tsfstr + '_' + str(s[3])

    for train_index, val_index in rskf.split(df_Test_Train.index, df_Test_Train.sex):
        print(test_name)
        if norm:
            data_fold = (ImageList.from_df(df_Test_Train, '/home/jplineh/Chicken_Proj')
                        .split_by_ids(train_index, val_index)
                        .label_from_df(cols='sex')
                        .transform(x[2], size=224)
                        .databunch(bs = 2)).normalize()

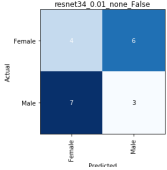
        else:
            data_fold = (ImageList.from_df(df_Test_Train, '/home/jplineh/Chicken_Proj')
                        .split_by_ids(train_index, val_index)
                        .label_from_df(cols='sex')
                        .transform(x[2], size=224)
                        .databunch(bs = 2))

        learn = cnn_Learner(data_fold, arch, metrics=error_rate, callback_fns = [CSVLogger], wd=wd)
        learn.fit_one_cycle(epoch_cycles)

        df_history = pd.read_csv('history.csv') # Appends to dataframe
        DFBig.append({'test_name': test_name,
                      'model_arch': archstr,
                      'transforms': tsfstr,
                      'normalized': str(norm),
                      'weight_decay': wd,
                      'split_num': split_num, # indicates hfold split
                      'max_error': df_history.error_rate.max(),
                      'min_error': df_history.error_rate.min(),
                      'avg_error': df_history.error_rate.mean(),
                      'train_df': df_history, ignore_index = True})

        interp = ClassificationInterpretation.from_learner(learn)
        interp.plot_confusion_matrix(return_fig=True, title=test_name)
        split_num += 1
        DFBig.to_csv('DF_resnet34_resnet50.csv')
```

/usr/local/lib/python3.6/dist-packages/matplotlib/pyplot.py:514: RuntimeWarning: More than 20 figures have been opened. Figures created through the pyplot interface ('matplotlib.pyplot.figure') are retained until explicitly closed and may consume too much memory. (To control this warning, see the rcParam 'figure.max_open_warning').
max_open_warning



Results from the testing:

Index	(min_error, 'average')	(min_error, '<lambda>')	(max_mean_error, ')
resnet34_0.003_modified_False	0.424722	0.0283548	0.481432
resnet34_0.01_none_False	0.409167	0.0290835	0.467334
resnet50_1_modified_True	0.369444	0.0243115	0.418067
resnet50_0.01_modified_True	0.360556	0.032988	0.426532

However we realized that using the min_error method of finding the best hyperparameters leaves in bias so we continued with AUROC from here on out.

[New hyperparameter method \(Previous%20best%20and%20new%20hyperparameter%20search.ipynb\)](#)