

```
In [1]: import os
from fastai import *
from fastai.vision import *
from fastai.callbacks import *
import numpy as np
import pandas as pd
pd.options.display.max_rows = 999
```

Modifying the the Test-Train CSV to work with fastai

```
In [2]: df_Test_Train = pd.read_csv('cropped_test_train.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 = df_Test_Train.drop(columns=['index'])
df_Test_Train.head(100)
```

19	./Cropped_Egg_images/Clutch1_D18Egg34.JPG	1	18	34	Female
20	./Cropped_Egg_images/Clutch1_D18Egg35.JPG	1	18	35	Male
21	./Cropped_Egg_images/Clutch1_D18Egg36.JPG	1	18	36	Male
22	./Cropped_Egg_images/Clutch1_D18Egg37.JPG	1	18	37	Female
23	./Cropped_Egg_images/Clutch1_D18Egg38.JPG	1	18	38	Male
24	./Cropped_Egg_images/Clutch1_D18Egg39.JPG	1	18	39	Male
25	./Cropped_Egg_images/Clutch1_D18Egg40.JPG	1	18	40	Female
26	./Cropped_Egg_images/Clutch1_D18Egg41.JPG	1	18	41	Male
27	./Cropped_Egg_images/Clutch1_D18Egg42.JPG	1	18	42	Male
28	./Cropped_Egg_images/Clutch1_D18Egg45.JPG	1	18	45	Female
29	./Cropped_Egg_images/Clutch1_D18Egg46.JPG	1	18	46	Male
30	./Cropped_Egg_images/Clutch1_D18Egg47.JPG	1	18	47	Male
31	./Cropped_Egg_images/Clutch1_D18Egg51.JPG	1	18	51	Male

```
In [3]: def dropdot(row):
    filepath = row['Crp_Filepath']
    newfilepath = filepath.split('/')[-1]
    return newfilepath

df_Test_Train['Crp_Filepath'] = df_Test_Train.apply(dropdot, axis =1)

#Get just the D18
df_Test_Train = df_Test_Train[df_Test_Train['Day']==18]
df_Test_Train = df_Test_Train.reset_index()
df_Test_Train = df_Test_Train.drop(columns = ['index'])
df_Test_Train.head()

df_Test_Train.to_csv('Fastai_dataset_usable_d14.csv')
```

```
In [4]: df_Test_Train.head(100)
```

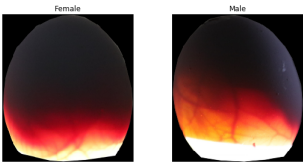
Out[4]:

	Crp_Filepath	clutch	Day	egg_number	sex
0	Cropped_Egg_images/Clutch2_D14IMG_0002.JPG	2	14	1	Female
1	Cropped_Egg_images/Clutch2_D14IMG_0003.JPG	2	14	2	Male
2	Cropped_Egg_images/Clutch2_D14IMG_0007.JPG	2	14	6	Female
3	Cropped_Egg_images/Clutch2_D14IMG_0008.JPG	2	14	7	Female
4	Cropped_Egg_images/Clutch2_D14IMG_0010.JPG	2	14	9	Male
5	Cropped_Egg_images/Clutch2_D14IMG_0013.JPG	2	14	12	Male
6	Cropped_Egg_images/Clutch2_D14IMG_0014.JPG	2	14	13	Female
7	Cropped_Egg_images/Clutch2_D14IMG_0016.JPG	2	14	15	Female
8	Cropped_Egg_images/Clutch2_D14IMG_0018.JPG	2	14	17	Female
9	Cropped_Egg_images/Clutch2_D14IMG_0022.JPG	2	14	21	Female
10	Cropped_Egg_images/Clutch2_D14IMG_0025.JPG	2	14	24	Female

Initial testing

```
In [165]: np.random.seed(42)
data = ImageDataBunch.from_df('/home/jplineb/Chicken_Proj', df_Test_Train, label_col = 'sex', size = (252,224), bs = 2, valid_pct=0.20, ds_tfms = None)
```

```
In [166]: data.show_batch()
```



```
In [164]: data.batch_stats
```

Out[164]: <bound method ImageDataBunch.batch_stats of ImageDataBunch>

Train: LabelList (76 items)
x: ImageList
Image (3, 224, 252),Image (3, 224, 252),Image (3, 224, 252),Image (3, 224, 252),Image (3, 224, 252)
y: CategoryList
Female,Male,Female,Female,Female
Path: /home/jplineb/Chicken_Proj;

Valid: LabelList (18 items)
x: ImageList
Image (3, 224, 252),Image (3, 224, 252),Image (3, 224, 252),Image (3, 224, 252),Image (3, 224, 252)
y: CategoryList
Female,Female,Female,Female,Female
Path: /home/jplineb/Chicken_Proj;

Test: None

```
In [174]: # Attempting to use CSVLogger callback
learn = cnn_learner(data, models.resnet34, metrics = error_rate, wd=10, callback_fns=[CSVLogger])
```

```
In [175]: learn.fit_one_cycle(6)
learn.recorder.plot_losses()
```

Total time: 00:16

epoch	train_loss	valid_loss	error_rate	time
0	0.868182	1.010944	0.611111	00:02
1	0.771957	0.717360	0.611111	00:02
2	0.722283	0.702474	0.722222	00:02
3	0.704669	0.696288	0.666667	00:02
4	0.697446	0.695262	0.666667	00:02
5	0.694409	0.695622	0.666667	00:02



```
In [176]: # CSVLogger callback testing
df_history = pd.read_csv('history.csv')
df_history.head()
```

Out[176]:

	epoch	train_loss	valid_loss	error_rate	time
0	0	0.868182	1.010944	0.611111	NaN
1	1	0.771957	0.717360	0.611111	NaN
2	2	0.722283	0.702474	0.722222	NaN
3	3	0.704669	0.696288	0.666667	NaN
4	4	0.697446	0.695262	0.666667	NaN

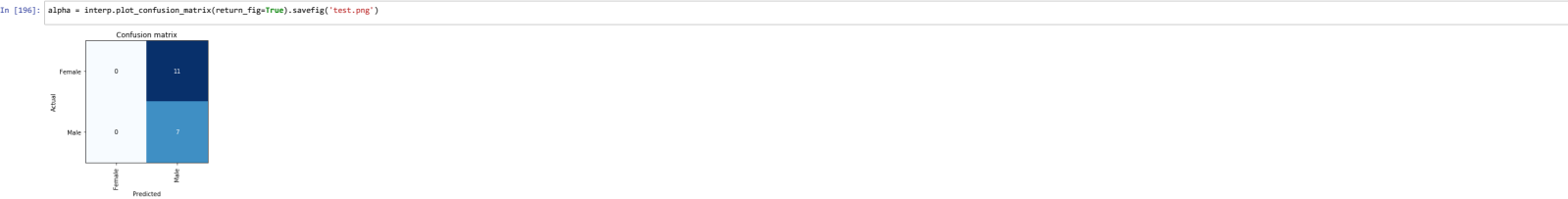
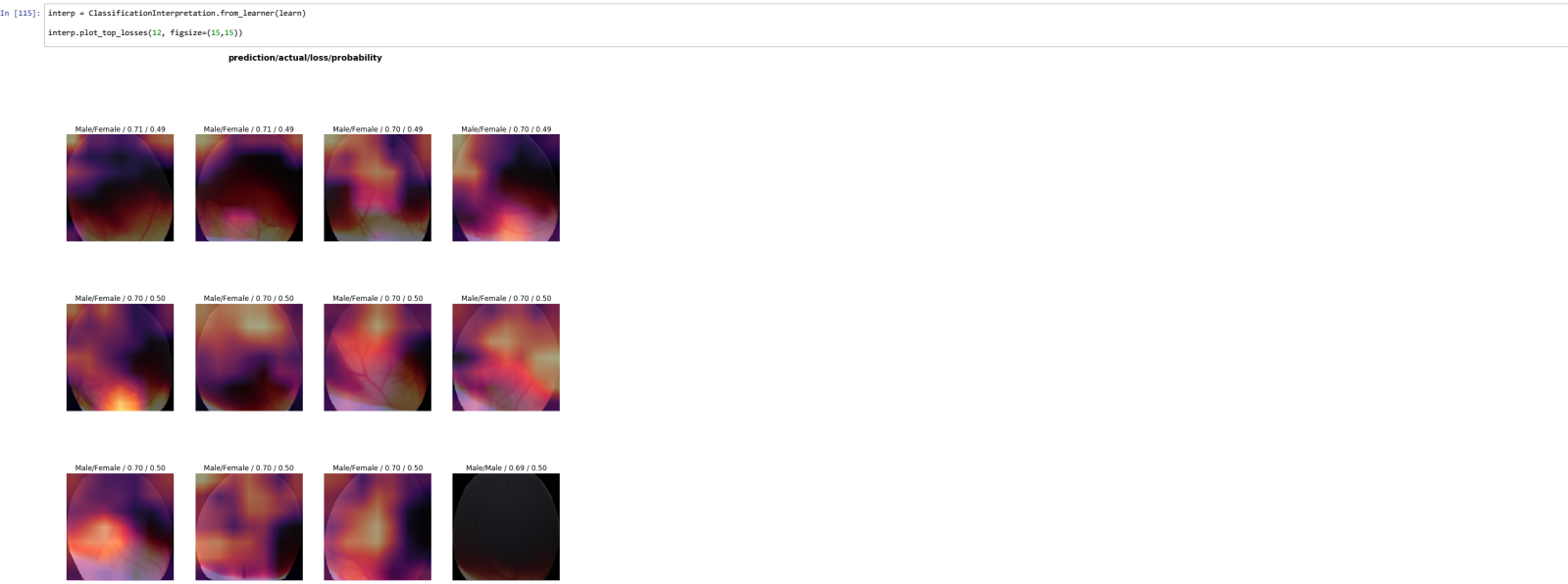
```
In [178]: df_history.error_rate.max()
```

```
Out[178]: 0.722222
```

```
In [191]: # Testing dataframe creation #
DFBig = pd.DataFrame(columns = ['model_arch', 'transforms', 'normalized', 'weight_decay', 'max_error', 'min_error', 'avg_error', 'train_df'])
DFBig = DFBig.append({'model_arch': 'resnet34',
    'transforms': 'None',
    'normalized': 'False',
    'weight_decay': int(10),
    '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})
```

Out[191]:

	epoch	train_loss	valid_loss	error_rate	time
0	0	0.868182	1.010944	0.611111	NaN
1	1	0.771957	0.717360	0.611111	NaN
2	2	0.722283	0.702474	0.722222	NaN
3	3	0.704669	0.696288	0.666667	NaN
4	4	0.697446	0.695262	0.666667	NaN
5	5	0.694409	0.695622	0.666667	NaN



For Loop

```
In [ ]: ### A Function to test multiple parameters for training f0st0i models ## Second Technique
...
Things we want to vary:
Model Architecture, Transforms, Normalization, Weight Decay
Things we want to store:
Training Results, lowest error, individual model(unfrozen), confusion matrix
...

# Parameters to vary
modelarch = [models.resnet18, models.resnet34, models.resnet50]
transforms = [None, get_transforms()]
normalization = [True, False]
weight_decay = [.01, .1, 1, 10, 20]
epoch_cycles = 7

# Creating Frame Work
DFBig = pd.DataFrame(columns = ['model_arch', 'transforms', 'normalized', 'weight_decay', 'max_error', 'min_error', 'avg_error', 'train_df'])
np.random.seed(42)
for trans in transforms:
    for norm in normalization:
        if norm:
            data = ImageDataBunch.from_df('/home/jplimb/Chicken_Proj', df_Test_Train, label_col = 'sex', size = 224, bs = 2, valid_pct=0.20, ds_tfms = trans).normalize()
        else:
            data = ImageDataBunch.from_df('/home/jplimb/Chicken_Proj', df_Test_Train, label_col = 'sex', size = 224, bs = 2, valid_pct=0.20, ds_tfms = trans)
        for arch in modelarch:
            archstr = str(arch).split(' ')[1]
            for wd in weight_decay:
                if trans is not None:
                    trans = 'stock'
                test_name = archstr + "_" + str(trans) + "." + str(norm) + "." + str(wd)
                learn = cnn_learner(data, arch, metrics = error_rate, wdwd, callback_fns=[CSVLogger])
                learn.fit_one_cycle(epoch_cycles)
                learn.save(test_name)
                df_history = pd.read_csv('history.csv')
                DFBig = DFBig.append({'model_arch': archstr,
                                     'transforms': str(trans),
                                     'normalized': str(norm),
                                     'weight_decay': wd,
                                     '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)#.savefig('./matrices/' + test_name + '.png')
            DFBig.to_csv('DF_Big.csv')
```

Results from this test found Resnet50 and Resnet34 needed further testing

df_agg_clean - DataFrame

Index	('min_error', 'average')	('min_error', 'clambda>')	('max_mean_error', ')
resnet50_0.01_modified_True	0.308772	0.0379857	0.384743
resnet50_1_modified_True	0.329825	0.0273594	0.384543
resnet34_0.003_modified_False	0.340351	0.0235899	0.387531
resnet34_0.01_None_False	0.340935	0.0140741	0.369084
resnet50_0.01_Stock_True	0.360819	0.0295145	0.419848
resnet34_0.1_modified_False	0.361403	0.0265371	0.414477
resnet18_5_None_False	0.361403	0.0450984	0.4516
resnet34_0.003_None_True	0.361403	0.0219686	0.405341
resnet50_5_None_True	0.361988	0.0483254	0.458639
resnet18_0.003_Stock_True	0.362573	0.0365017	0.435576
resnet18_0.01_Stock_False	0.362573	0.0515948	0.465763
resnet18_0.1_Stock_False	0.362573	0.025837	0.414247
resnet34_5_None_True	0.362573	0.0394205	0.441414
resnet34_0.01_modified_True	0.362573	0.0365017	0.435577
resnet50_0.003_Stock_True	0.37076	0.0523214	0.475403
resnet34_1_modified_False	0.371345	0.0378811	0.447107
resnet18_1_Stock_False	0.374269	0.0415494	0.457368
resnet50_0.01_None_True	0.382456	0.0334348	0.449325
resnet50_1_Stock_True	0.393041	0.0544583	0.491957
resnet34_0.1_None_False	0.383041	0.017505	0.418051

Format Resize ☒ Background color ☒ Column min/max Save and Close Close

For Further testing of these parameters look at: [Repeated Stratified K Fold on "Good" hyperparameters \(Repeated%20Stratified%20Kfold.ipynb\)](#)

```
[15]: ## Test of Cross Validation Loading
# This was used for Learning how to use sklearn's stratified KFold
from sklearn.model_selection import KFold
tfms = get_transforms()
kf = KFold(n_splits=2)
acc_val = []
kappa = KappaScore()
kappa.weights = "quadratic"
for train_index, val_index in kf.split(df_Test_Train.index):
    print(train_index, val_index)
    data_fold = (ImageList.from_df(df_Test_Train, '/home/jplineb/Chicken_Proj')
                  .split_by_idxs(train_index, val_index)
                  .label_from_df(cols='sex')
                  .transform(tfms, size=224)
                  .databunch(bs=2)).normalize()
    learn = cnn_learner(data_fold, models.resnet18, metrics=error_rate, wd=.1)
    learn.fit_one_cycle(7)
    loss, acc = learn.validate()
    acc_val.append(acc.numpy())
```

```
In [7]: from sklearn.metrics import roc_auc_score

def auroc_score(input, target):
    input, target = input.cpu().numpy()[1:], target.cpu().numpy()
    return roc_auc_score(target, input)

class AUROC(Callback):
    _order = -20 #needs to run before the recorder

    def __init__(self, learn, extra=None, **kwargs):
        self.learn = learn

    def on_train_begin(self, **kwargs):
        self.learn.recorder.add_metric_names(['AUROC'])

    def on_epoch_begin(self, **kwargs):
        self.output, self.target = [], []

    def on_batch_end(self, last_target, last_output, train, **kwargs):
        if not train:
            self.output.append(last_output)
            self.target.append(last_target)

    def on_epoch_end(self, last_metrics, **kwargs):
        if len(self.output) > 0:
            output = torch.cat(self.output)
            target = torch.cat(self.target)
            preds = F.softmax(output, dim=-1)
            metric = auroc_score(preds, target)
            return add_metrics(last_metrics, [metric])
```

```
In [ ]: # Iterative method with Cross Validation
from sklearn.model_selection import StratifiedKFold

# Parameters to vary
modelarch = [models.resnet18, models.resnet34, models.resnet50]
transforms = [None, get_transforms()],
get_transforms(do_flip = True, flip_vert = False, max_rotate = 35, max_lighting = None, max_warp = .2, p_lighting = 0)]
normalization = [True, False]
weight_decay = [.01, .1, .3, .5, 1]
epoch_cycles = 9

# Creating from work
DFBig = pd.DataFrame(columns = ['test_name', 'model_arch', 'transforms', 'normalized',
                                'weight_decay', 'split_num'])
np.random.seed(42) #locks random seed
skf = StratifiedKFold(n_splits=5)
for arch in modelarch:
    archstr = str(arch).split(' ')[1]
    for wd in weight_decay:
        #kf = KFold(n_splits=5) # creates k fold crossvalidation with n splits
        #acc_val = [] # create an empty list to store accuracy
        #test_name = archstr + '_' + 'tsf' + '_' + 'norm' + '_' + str(wd) # this run locks in the transforms and normalization; that variability comes in in later tests

        for norm in normalization:
            for tsfns in transforms:
                if tsfns is not None:
                    if tsfns == get_transforms():
                        tsfstr = 'Stock'
                    else:
                        tsfstr = 'modified'
                    tsfstr = 'None'

                test_name = archstr + '_' + str(wd) + '_' + tsfstr + '_' + str(norm)
                print(test_name)
                split_num = 1
                #for train_index, val_index in kf.split(df_Test_Train.index):
                for train_index, val_index in skf.split(df_Test_Train.index, df_Test_Train['sex']):
                    if norm:
                        data_fold = (ImageList.from_df(df_Test_Train, '/home/jplineb/Chicken_Proj')
                                      .split_by_idxs(train_index, val_index)
                                      .label_from_df(cols='sex')
                                      .transform(tsfns, size=224)
                                      .databunch(bs=2)).normalize()
                    else:
                        data_fold = (ImageList.from_df(df_Test_Train, '/home/jplineb/Chicken_Proj')
                                      .split_by_idxs(train_index, val_index)
                                      .label_from_df(cols='sex')
                                      .transform(tsfns, size=224)
                                      .databunch(bs=2))

                learn = cnn_learner(data_fold, arch, metrics=error_rate, callback_fns = [CVLogger, AUROC, partial(EarlyStoppingCallback, monitor='AUROC', mode='max', min_delta=0.01, patience=100)], wd=wd)
                learn.fit_one_cycle(epoch_cycles)

                df_history = pd.read_csv('history.csv') # adds to dataframe created earlier (again norm and tsfns locked to 0N)
                DFBig = DFBig.append({'test_name': test_name,
                                      'model_arch': archstr,
                                      'transforms': tsfstr,
                                      'normalized': str(norm),
                                      'weight_decay': wd,
                                      'split_num': split_num, # indicates kfold split
                                      'error_rate_0': df_history.error_rate[7],
                                      'error_rate_9': df_history.error_rate[8],
                                      'AUROC_0': df_history.AUROC[7],
                                      'AUROC_9': df_history.AUROC[8],
                                      'train_df': df_history, ignore_index = True})

                split_num+=1
            #interp = ClassificationInterpretation.from_learner(learn)
            #interp.plot_confusion_matrix(return_fig=True, title=test_name)#savefig('./matrices/' + test_name + '.png')
            DFBig.to_csv('DF_Big.csv')
```

resnet18_0.01_None_True

Total Time: 00:13

epoch	train_loss	valid_loss	error_rate	AUROC	time
0	1.091993	0.868008	0.750000	0.437500	00:01
1	0.938231	0.846525	0.625000	0.375000	00:01
2	0.937822	1.076271	0.625000	0.437500	00:01
3	0.834414	0.925072	0.250000	0.625000	00:01
4	0.836165	0.761775	0.375000	0.625000	00:01
5	0.752544	0.844406	0.500000	0.625000	00:01
6	0.708230	0.838999	0.500000	0.625000	00:01
7	0.707960	0.933295	0.375000	0.625000	00:01
8	0.751757	0.764272	0.500000	0.625000	00:01

Results from the testing

df_mean - DataFrame

	Index	Unnamed: 0	weight_decay	split_num	AUROC_8	AUROC_9	error_rate_8	error_rate_9
	resnet18_0.01_None_False	17	0.01	3	0.670833	0.620833	0.410714	0.382143
	resnet50_0.3_None_True	362	0.3	3	0.654167	0.666667	0.360714	0.385714
	resnet50_0.5_modified_T...	399.5	0.5	3	0.652083	0.583333	0.451786	0.435714
	resnet50_0.01_None_False	317	0.01	3	0.633333	0.6875	0.407143	0.407143
	resnet50_1_None_False	437	1	3	0.608333	0.6	0.442857	0.439286
	resnet50_0.3_None_False	377	0.3	3	0.591667	0.570833	0.564286	0.539286
	resnet50_0.01_modified_...	324.5	0.01	3	0.5875	0.597917	0.430357	0.455357
	resnet18_1_modified_Fal...	144.5	1	3	0.5875	0.591667	0.425	0.446429
	resnet50_0.5_modified_F...	414.5	0.5	3	0.575	0.591667	0.45	0.5
	resnet18_0.01_Stock_True	7	0.01	3	0.566667	0.5375	0.485714	0.435714
	resnet34_0.1_None_False	197	0.1	3	0.558333	0.558333	0.457143	0.432143
	resnet18_0.3_modified_F...	84.5	0.3	3	0.55625	0.55625	0.476786	0.4625
	resnet50_0.1_None_False	347	0.1	3	0.55	0.608333	0.485714	0.485714
	resnet50_0.5_None_True	392	0.5	3	0.541667	0.525	0.439286	0.489286
	resnet18_0.5_modified_T...	99.5	0.5	3	0.5375	0.539583	0.491071	0.491071
	resnet18_1_None_True	122	1	3	0.533333	0.545833	0.460714	0.464286
	resnet34_0.01_None_True	152	0.01	3	0.533333	0.5375	0.592857	0.485714
	resnet34_0.01_None_False	167	0.01	3	0.529167	0.583333	0.535714	0.535714
	resnet50_0.5_None_False	407	0.5	3	0.529167	0.529167	0.407143	0.435714
	resnet34_0.1_modified_T...	189.5	0.1	3	0.525	0.491667	0.553571	0.501786

Format

Resize

☒ Background color

☒ Column min/max

Save and Close

Close

Best Results

- Resnet18
- No Transforms
- wd = 0.01
- 9 epochs
- normalization off
- max lr not set

[Search for Repeated Stratified K-Fold on best results \(Previous%20best%20and%20new%20hyperparameter%20search.ipynb\)](#)

Devils Advocate Testing

Hypothesis: Flip day 1 gender

```
In [137]: def changegend(row):
         index = row.name
         gend = row['sex']
         if index < 55:
             if gend=='Female':
                 newgend = 'Male'
             if gend=='Male':
                 newgend = 'Female'
         else:
             newgend = gend
         return (newgend)

df_swapped = pd.DataFrame(df_Test_Train)
df_swapped['sex'] = df_swapped.apply(changegend, axis = 1)
df_swapped.head(100)
```

Out[137]:

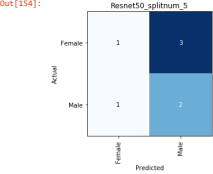
	Crp_Filepath	clutch	Day	egg_number	sex
0	Cropped_Egg_images/Clutch1_D18/egg2.JPG	1	18	2	Male
1	Cropped_Egg_images/Clutch1_D18/egg3.JPG	1	18	3	Male
2	Cropped_Egg_images/Clutch1_D18/egg4.JPG	1	18	4	Female
3	Cropped_Egg_images/Clutch1_D18/egg6.JPG	1	18	6	Male
4	Cropped_Egg_images/Clutch1_D18/egg9.JPG	1	18	9	Male
5	Cropped_Egg_images/Clutch1_D18/egg11.JPG	1	18	11	Male
6	Cropped_Egg_images/Clutch1_D18/egg15.JPG	1	18	15	Male
7	Cropped_Egg_images/Clutch1_D18/egg16.JPG	1	18	16	Female
8	Cropped_Egg_images/Clutch1_D18/egg17.JPG	1	18	17	Male
9	Cropped_Egg_images/Clutch1_D18/egg19.JPG	1	18	19	Female
10	Cropped_Egg_images/Clutch1_D18/egg20.JPG	1	18	20	Female

```
In [154]: from IPython.core.interactiveshell import InteractiveShell
         InteractiveShell.ast_node_interactivity = "all"

         import numpy as np
         from sklearn.model_selection import StratifiedFold
         np.random.seed(42)
         tfms = get_transforms(do_flip = True, flip_vert = False, max_rotate = 35, max_lighting = None, max_warp = .2, p_lighting = 0)
         kf = KFold(n_splits=5)
         skkf = StratifiedKFold(n_splits=5, shuffle = False)

         acc_val = []
         split_num = 1
         ## Attempt to use Just Clutch2 Data
         df_Test_Train_Clutch2 = df_Test_Train[55:]
         #df_Test_Train_Clutch2 = df_Test_Train[55:]
         #for train_index, val_index in skkf.split(df_Test_Train.index, df_Test_Train['sex']):
         #for train_index, val_index in kf.split(df_Test_Train.index):
         #for train_index, val_index in kf.split(df_Test_Train_Clutch2.index):
         for train_index, val_index in kf.split(df_swapped.index):
             print(train_index, val_index)
             test_name = ("Resnet18_splitnum_" + str(split_num))
             data_fold = (ImageList.from_df(df_Test_Train, "/home/jplineb/Chicken_Proj")
                           .split_by_index(train_index, val_index)
                           .label_from_df(cols='sex')
                           .transform(tfms, size=224)
                           .databunch(bs = 25)).normalize()

             learn = cnn_learner(data_fold, models.resnet18, metrics=error_rate, wd=.01)
             learn.save(test_name)
             learn.fit_one_cycle(7)
             loss, acc = learn.validate()
             learn.show_results()
             jinterp = ClassificationInterpretation.from_learner(learn)
             interp.plot_confusion_matrix(return_fig=True, title=test_name)
             acc_val.append(acc.numpy())
             split_num +=1
```

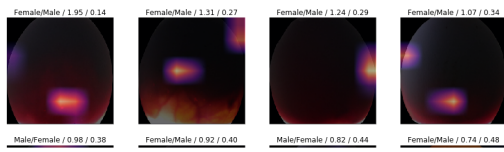


```
In [152]: print('Average error rate: ' + str(1-np.average(acc_val)))
         print('Standard error bounds: ' + str(np.std(acc_val)/np.sqrt(5)))

         Average error rate:0.44582925872802734
         Standard error bounds:0.84633143165237985
```

```
In [61]: interp = ClassificationInterpretation.from_learner(learn.load("Resnet50_splitrun_5"))
interp.plot_top_losses(16, figsize=(15,15))
```

prediction/actual/loss/probability



K-Fold Stratified Shuffle