

```
In [1]: %reload_ext autoreload
%autoreload 2
%matplotlib inline

In [2]: import torch

In [3]: torch.cuda.is_available()

Out[3]: True

In [4]: import os
from tqdm import tqdm, tnrange, tqdm_notebook
from pathlib import Path
import re
import numpy as np
import matplotlib.pyplot as plt
# import cv2
import sys
import scipy.ndimage
# from mpl_toolkits.mplot3d.art3d import Poly3DCollection

In [5]: #import pydicom
#from pydicom.data import get_testdata_files
#from pydicom.filereader import read_dicomdir
#import pydicom.pixel_data_handlers.gdcm_handler as gdcm_handler
# ! gdcm must be installed with conda install (conda install -c conda-forge gdcm)
# pydicom.config.image_handlers = ['gdcm_handler']

In [6]: # import nibabel as nib

In [7]: from fastai.vision import *
from fastai.metrics import *
from fastai.callbacks import *

In [8]: #from fastai2.data.all import *
#from fastai2.vision.core import *

In [9]: import pandas as pd
```

## Define paths

```
In [10]: path_str = '/home/ubuntu/sfr-challenge/lungs/dataset'
#path_str = '/Users/igorgarbz/SoftDev/sfr-challenge/dataset'

In [11]: path = Path(path_str)

In [12]: path_seg = path/'seg_3d'

In [13]: path_p = path/'Pathologiques'

In [14]: path_n = path/'Normaux'

In [15]: path_train = path_str + '/train'

In [16]: test_path = path_str + '/Pathologiques/N7Q0jai/N7Q0jai'
```

## Define fixed random seed

```
In [17]: np.random.seed(42)
```

## Test section ==>

```
In [18]: # cell to run the experiments
```

## <== End of test section

## Train network

```
In [19]: bs = 32  
valid_split = 0.2
```

```
In [20]: data = ImageDataBunch.from_folder(path='train', ds_tfms=get_transforms(), size=224, bs=bs, valid_pc
```

```
In [21]: pd.value_counts(data.train_dl.y.items.flatten(), sort=False)
```

```
Out[21]: 0    2672  
1     716  
dtype: int64
```

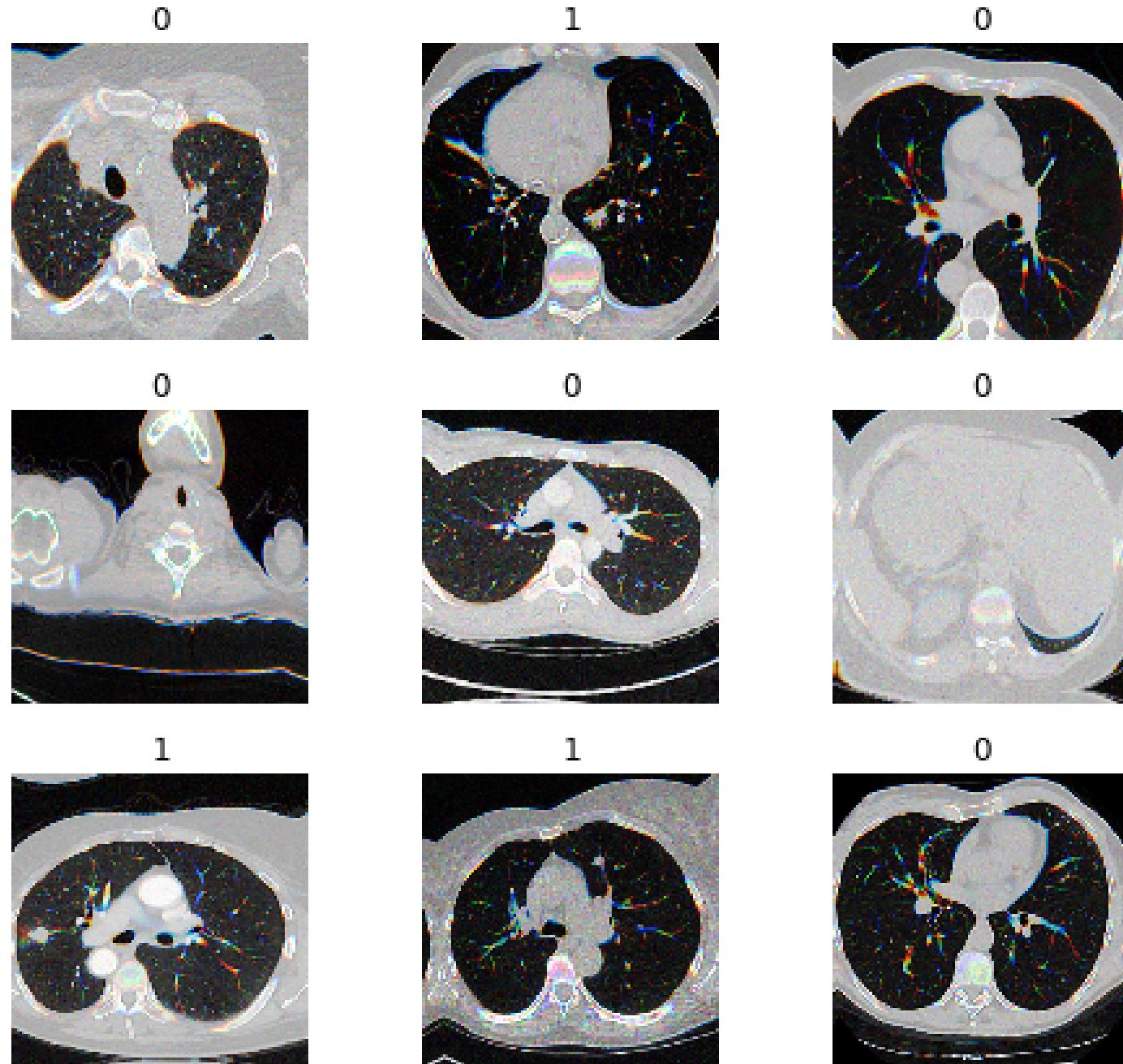
```
In [22]: pd.value_counts(data.valid_dl.y.items.flatten(), sort=False)
```

```
Out[22]: 0    669  
1     177  
dtype: int64
```

```
In [23]: data.classes
```

```
Out[23]: ['0', '1']
```

```
In [24]: data.show_batch(rows=3, figsize=(7,6))
```

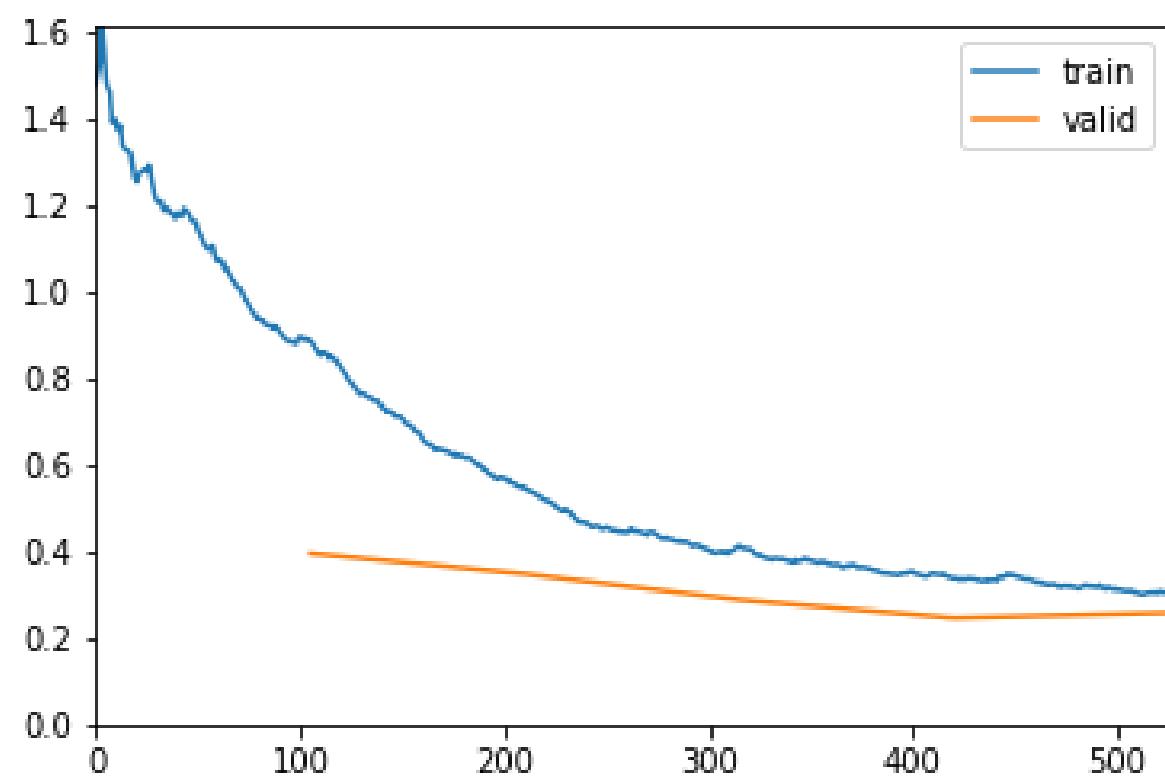


```
In [26]: learner = cnn_learner(data, models.vgg16_bn, metrics=[error_rate, AUROC()], callback_fns=[ShowGraph]  
#learner = cnn_learner(data, models.resnet18, metrics=[error_rate, f1_score(), AUROC()], callback_f
```

```
In [ ]: #verify_images(path='train')/0')
```

```
In [27]: learner.fit_one_cycle(5)
```

epoch	train_loss	valid_loss	error_rate	auroc	time
0	0.889512	0.394543	0.160756	0.862359	00:18
1	0.548829	0.347096	0.138298	0.868076	00:16
2	0.409311	0.287748	0.118203	0.915233	00:15
3	0.337301	0.245484	0.096927	0.940222	00:16
4	0.301509	0.257633	0.105201	0.935928	00:15



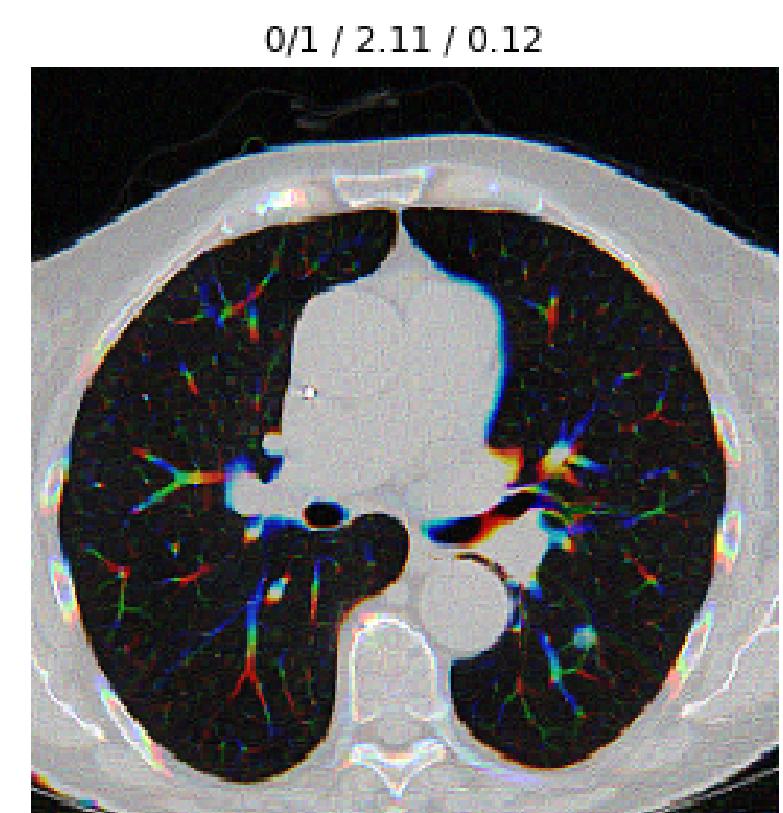
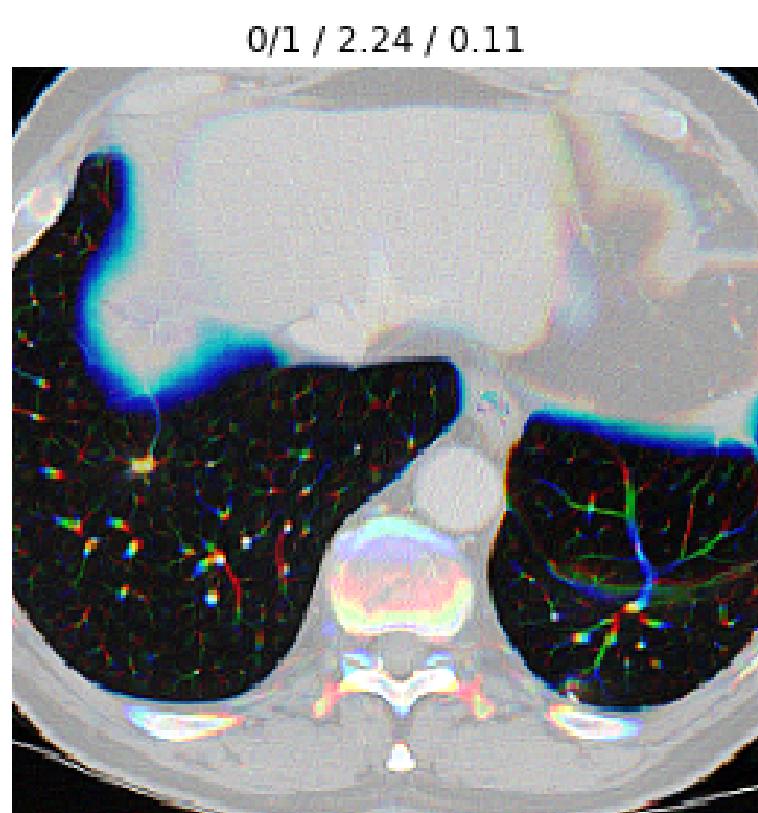
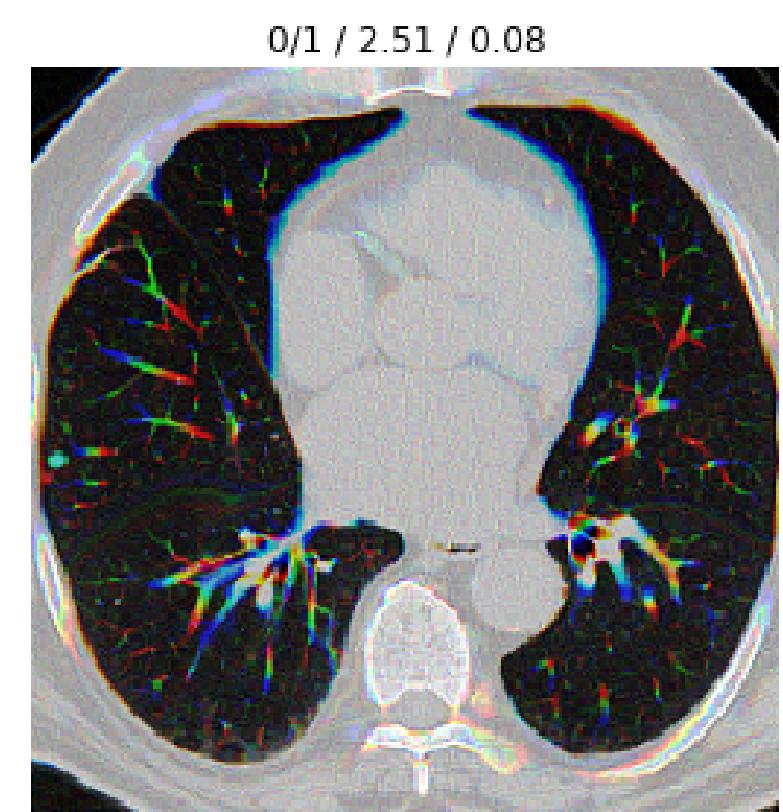
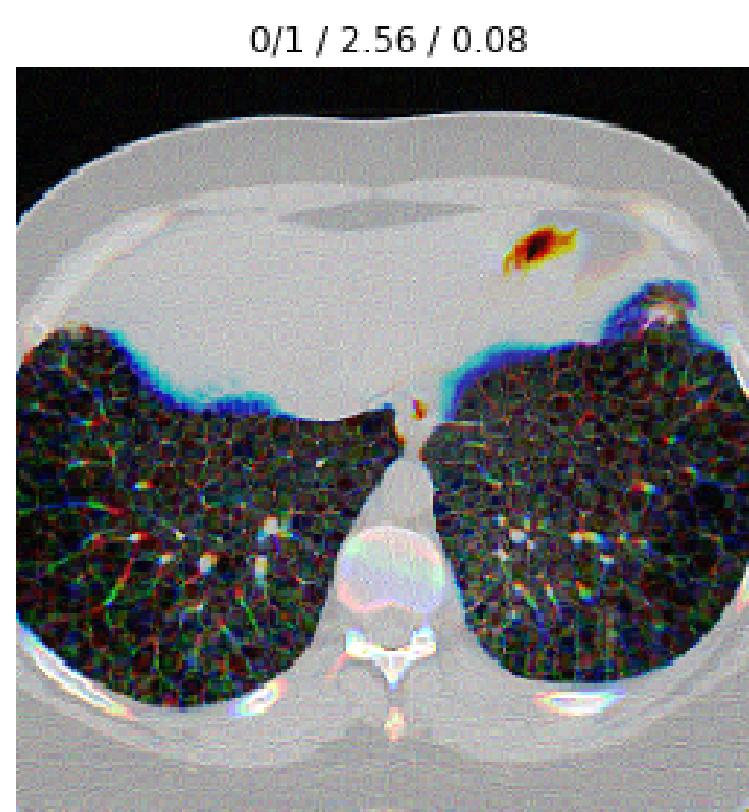
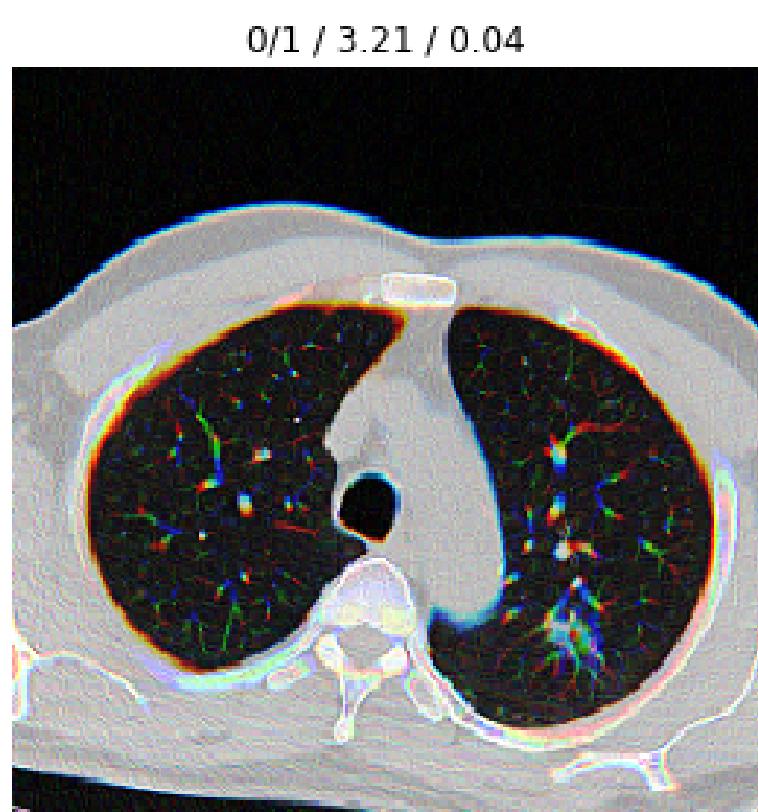
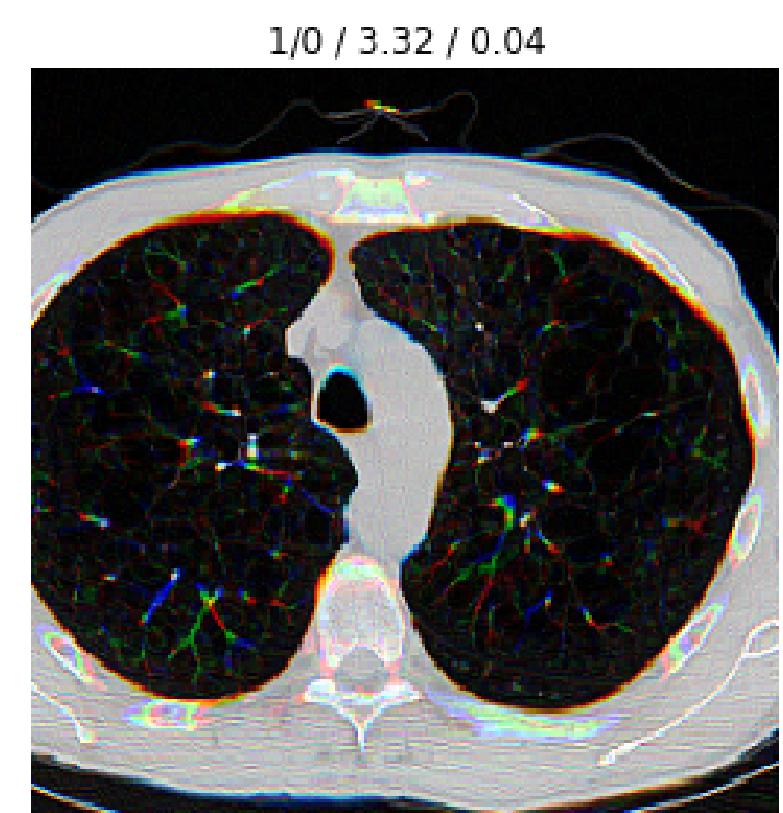
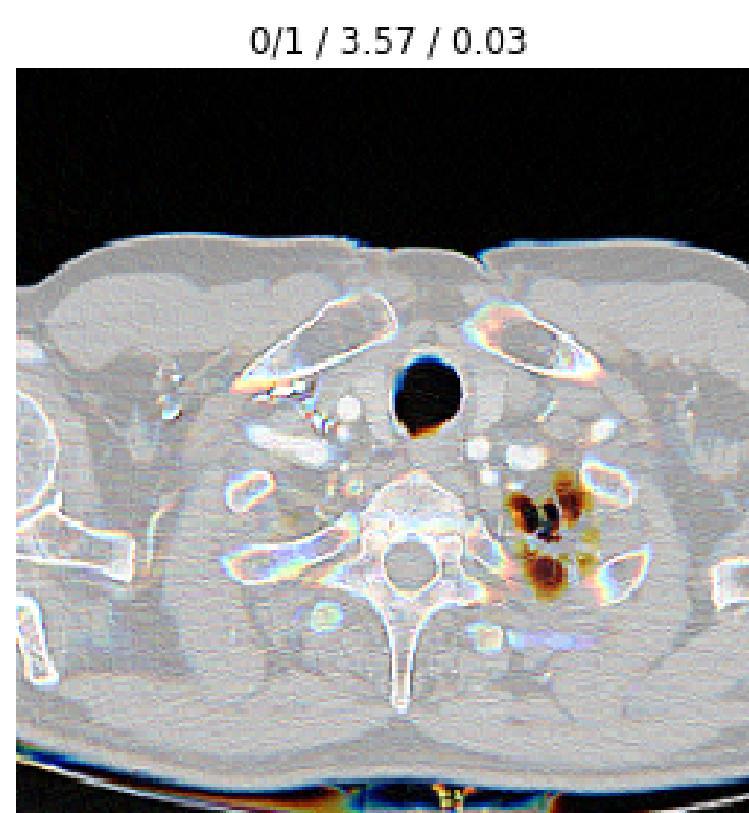
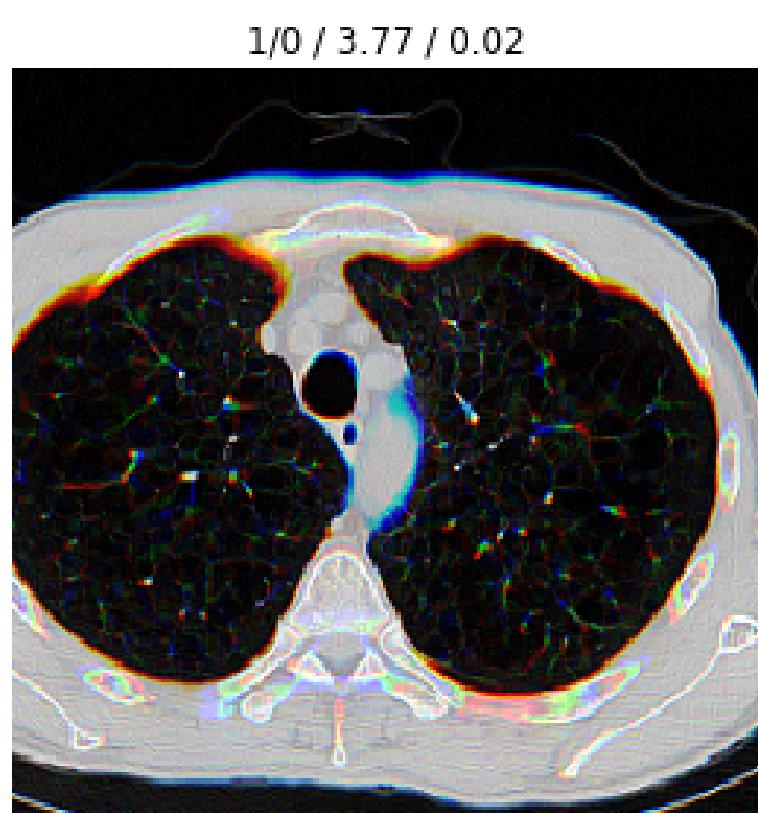
```
In [28]: learner.save('stage-1')
```

```
In [29]: interp = ClassificationInterpretation.from_learner(learner)
losses, idxs = interp.top_losses()
len(data.valid_ds)==len(losses)==len(idxs)
```

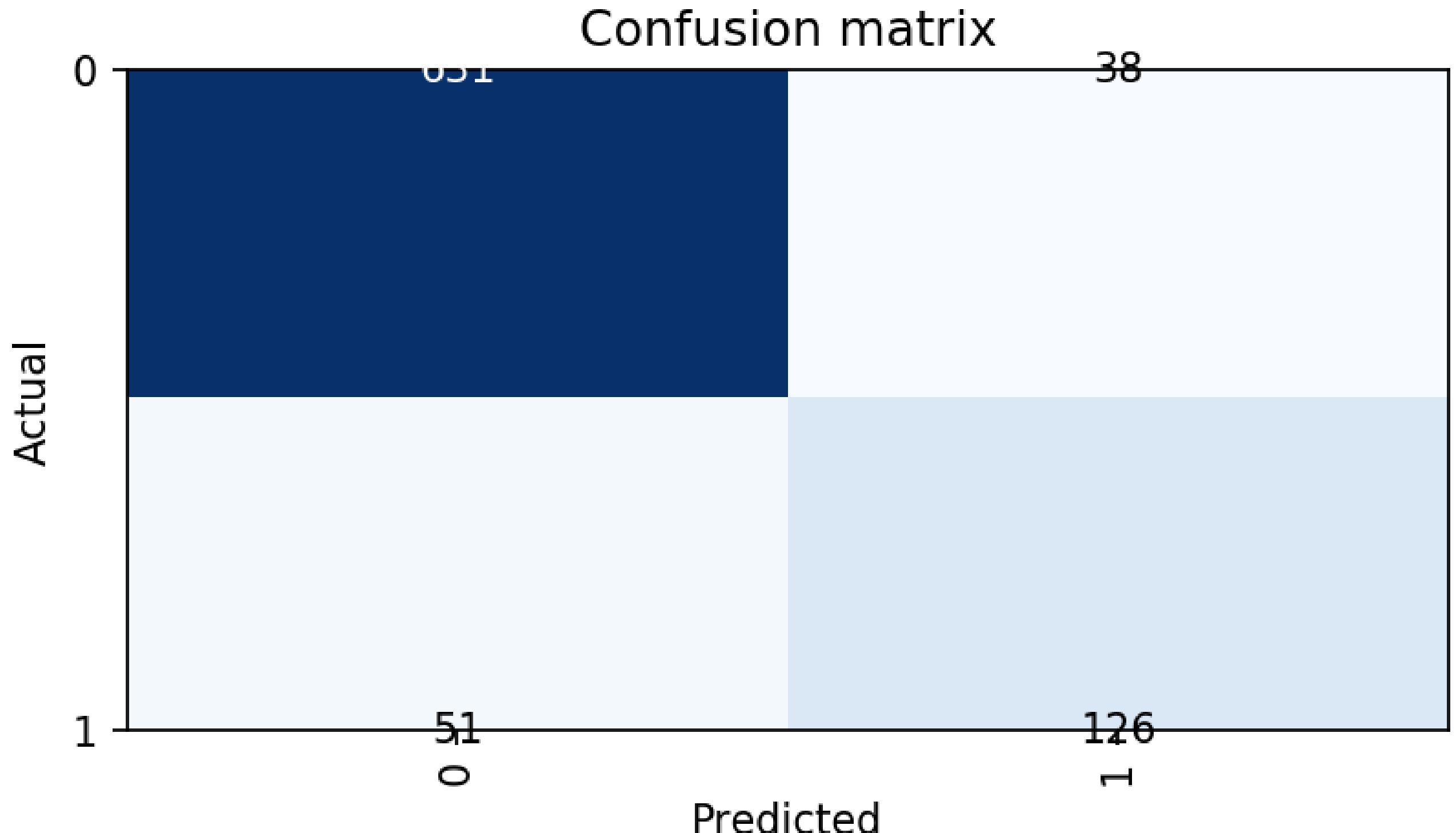
```
Out[29]: True
```

```
In [30]: interp.plot_top_losses(9, figsize=(16,16))
```

**prediction/actual/loss/probability**



```
In [31]: interp.plot_confusion_matrix(figsize=(5,5), dpi=160)
```



```
In [32]: interp.most_confused(min_val=2)
```

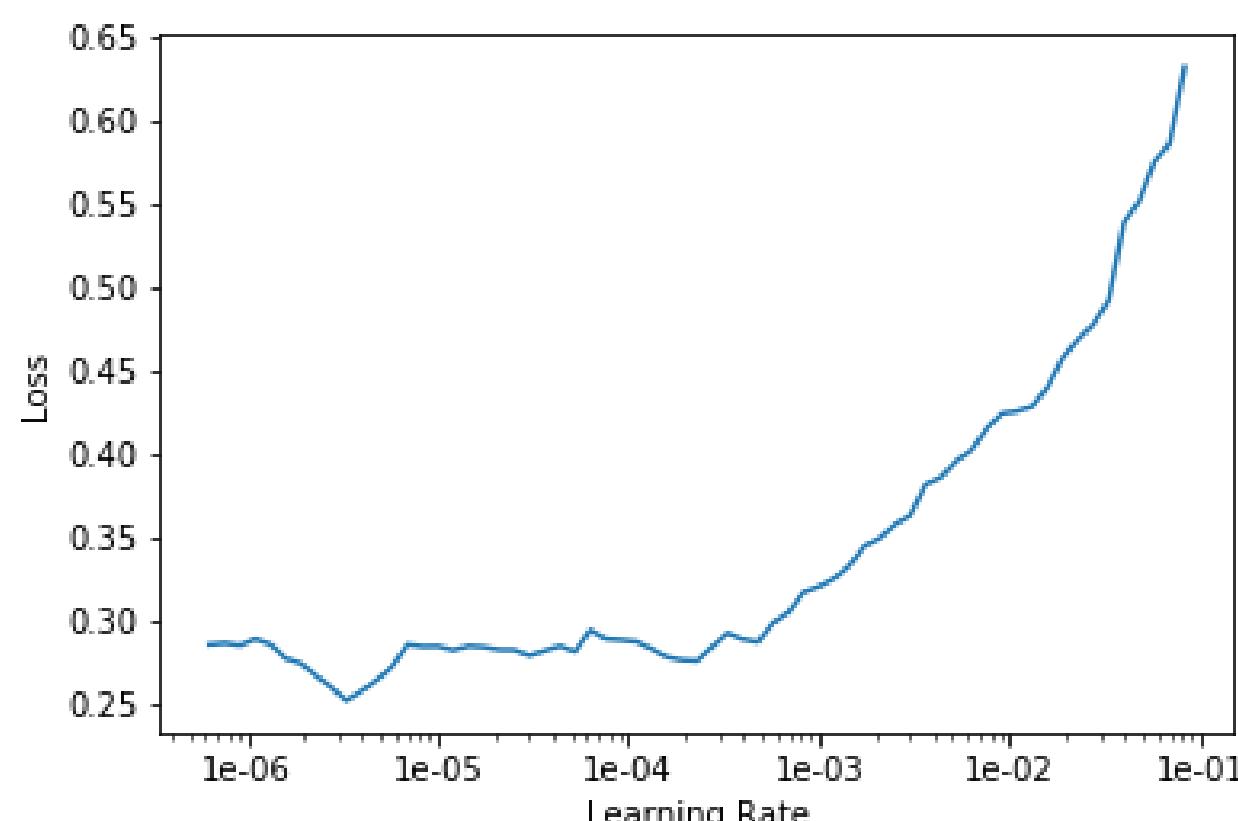
```
Out[32]: [('1', '0', 51), ('0', '1', 38)]
```

## Unfreeze all model layer and tune learning rate

```
In [33]: learner.unfreeze()  
learner.lr_find()
```

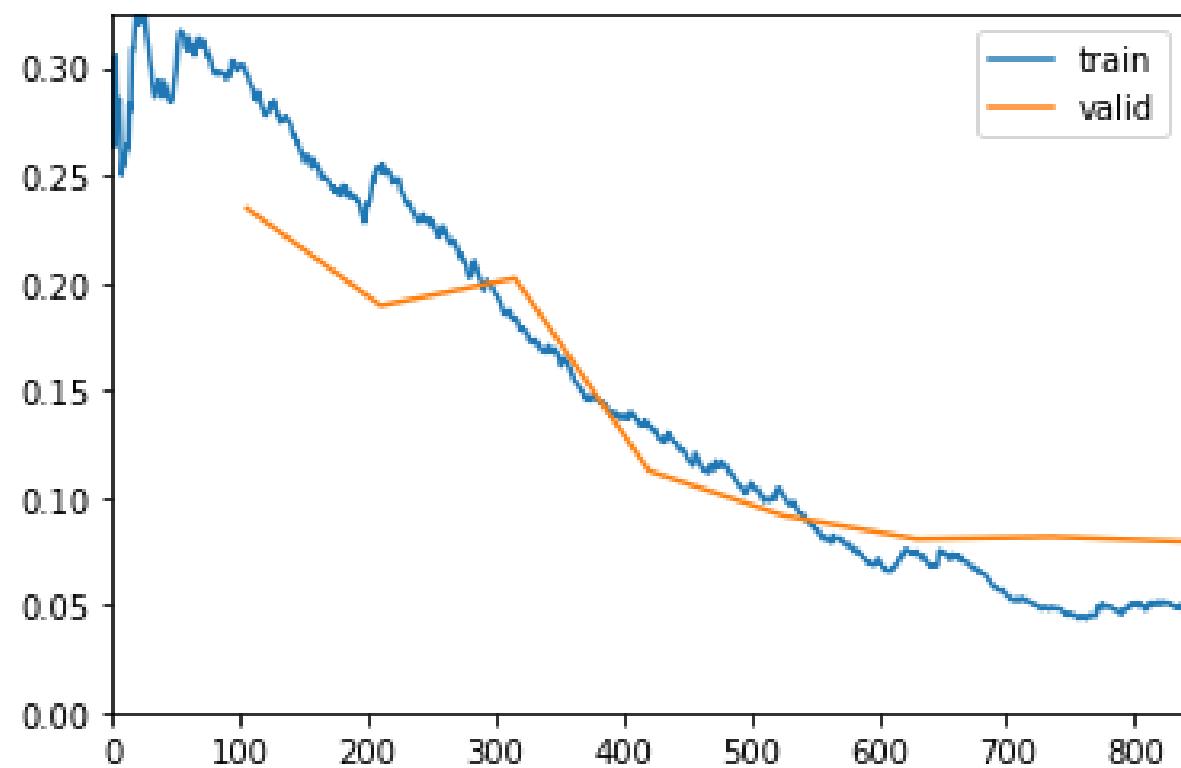
LR Finder is complete, type {learner\_name}.recorder.plot() to see the graph.

```
In [34]: learner.recorder.plot()
```



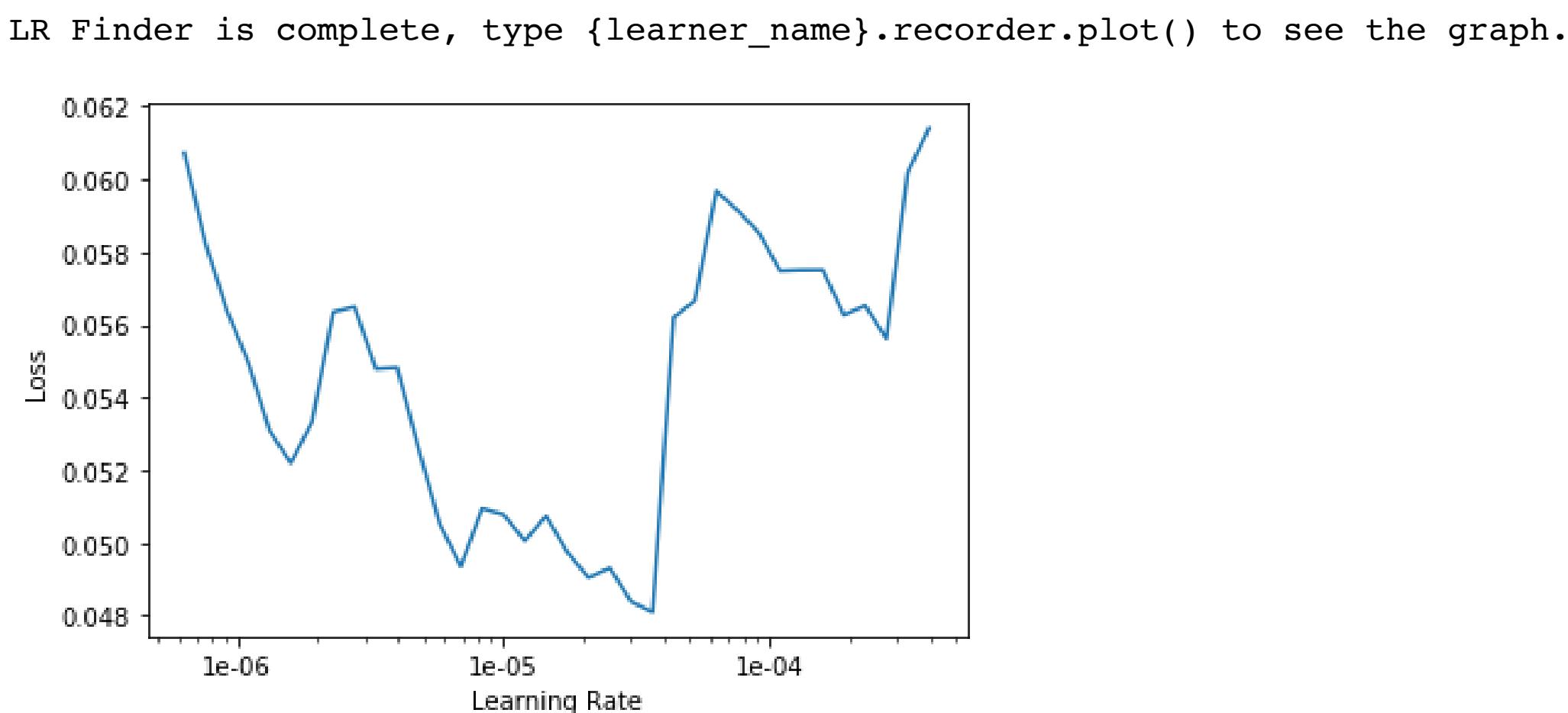
```
In [35]: learner.fit_one_cycle(8, max_lr=slice(1e-5, 3e-4))
```

epoch	train_loss	valid_loss	error_rate	auroc	time
0	0.297720	0.234594	0.100473	0.953130	00:17
1	0.252415	0.189253	0.072104	0.972098	00:17
2	0.183706	0.202034	0.081560	0.979229	00:17
3	0.133181	0.112315	0.034279	0.984596	00:17
4	0.101831	0.091629	0.030733	0.988164	00:18
5	0.075065	0.080705	0.021277	0.989727	00:18
6	0.048595	0.081809	0.027187	0.988836	00:17
7	0.051993	0.079529	0.026005	0.989752	00:17



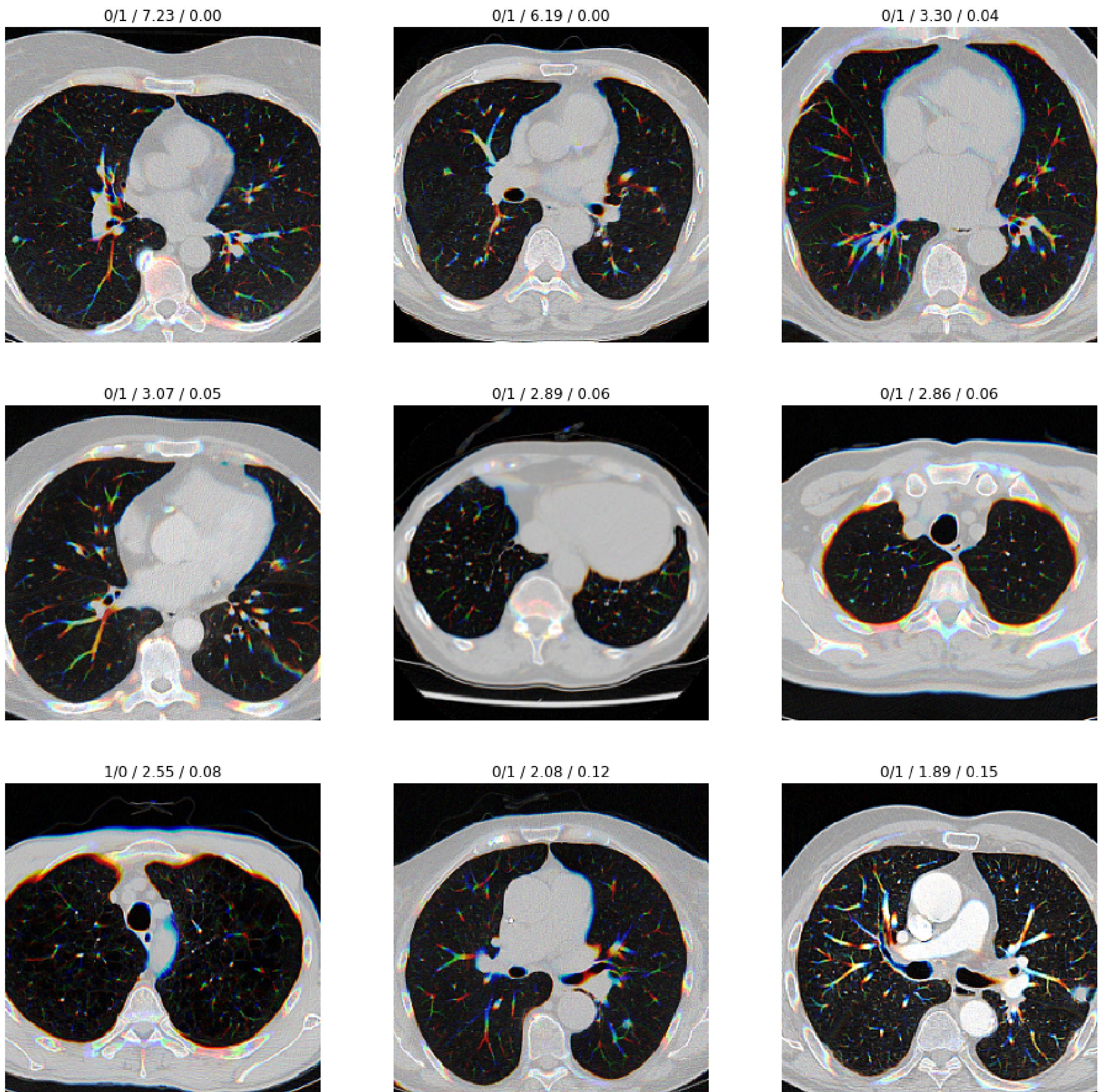
```
In [36]: learner.save('stage-2-8epc')
```

```
In [37]: learner.lr_find()
learner.recorder.plot()
```

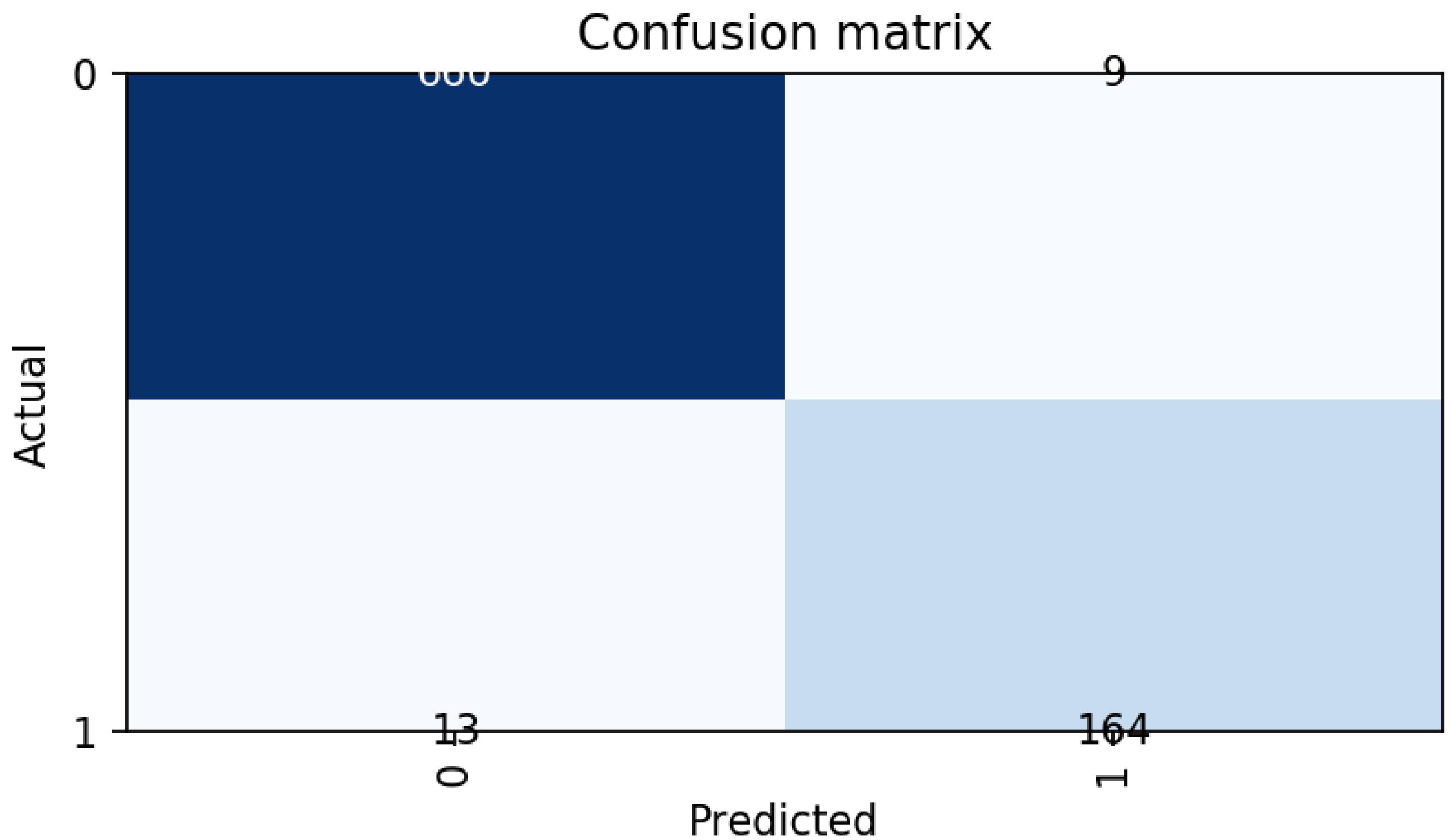


```
In [38]: interp = ClassificationInterpretation.from_learner(learner)
losses,idxs = interp.top_losses()
len(data.valid_ds)==len(losses)==len(idxs)
# plot
interp.plot_top_losses(9, figsize=(16,16))
```

**prediction/actual/loss/probability**



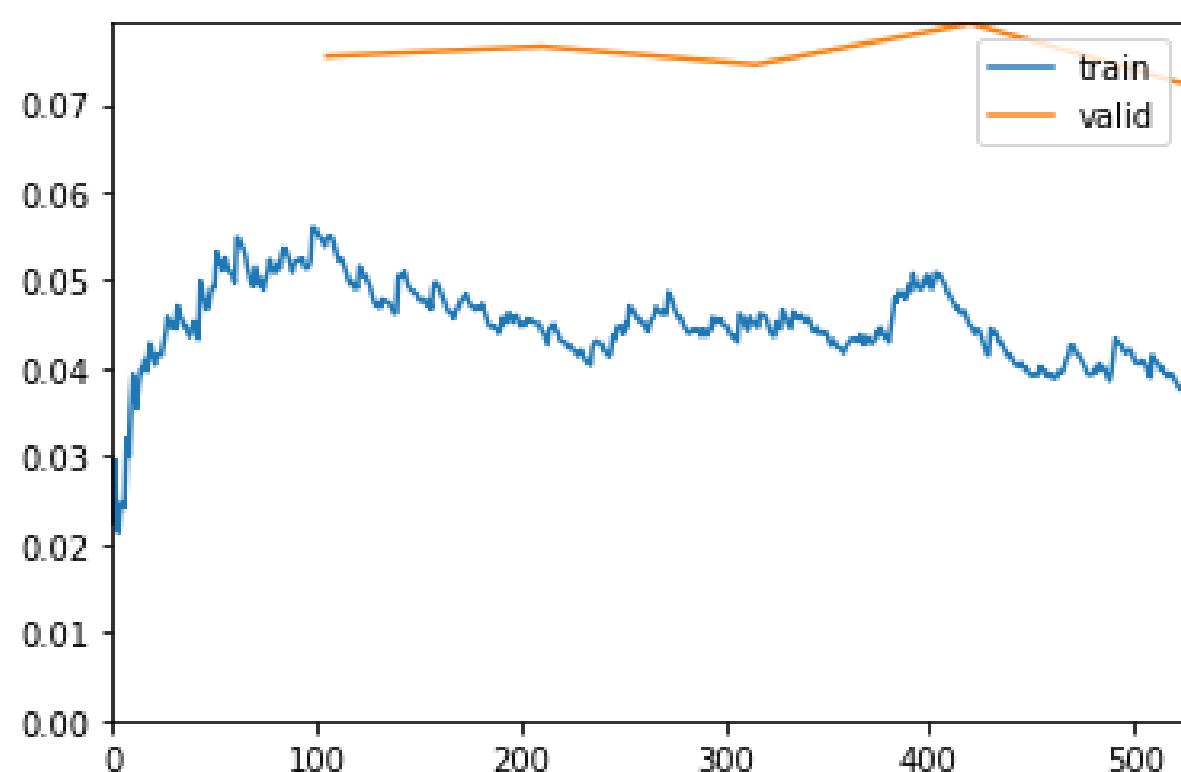
```
In [39]: interp.plot_confusion_matrix(figsize=(5,5), dpi=160)
```



Add x4 epochs using lower learning rate

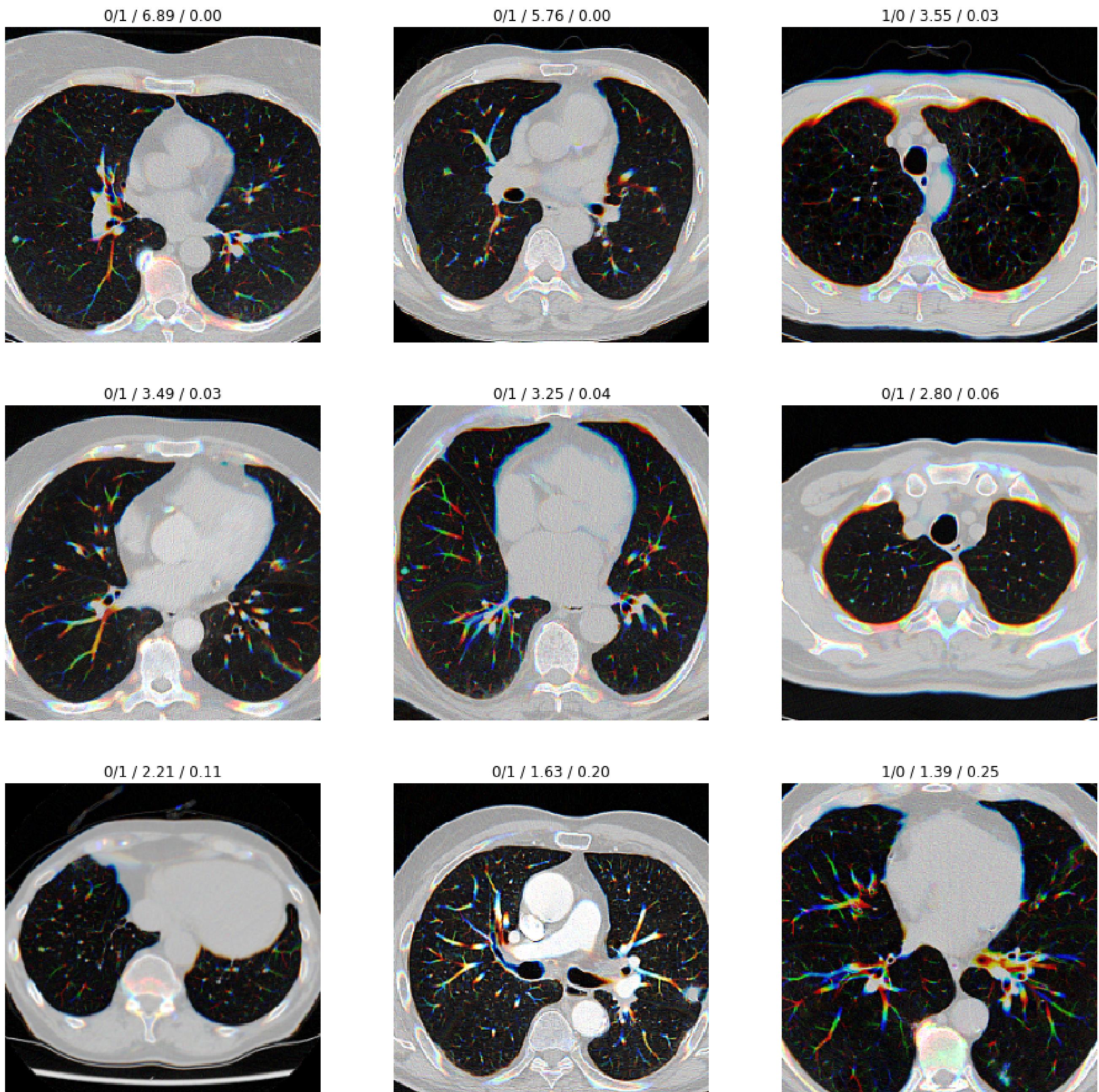
```
In [40]: learner.unfreeze()  
learner.fit_one_cycle(5, max_lr=1e-5)
```

epoch	train_loss	valid_loss	error_rate	auroc	time
0	0.053889	0.075421	0.018913	0.989140	00:17
1	0.045037	0.076536	0.026005	0.989292	00:18
2	0.045115	0.074487	0.024823	0.989659	00:18
3	0.045150	0.079163	0.026005	0.989731	00:17
4	0.037624	0.072283	0.026005	0.990390	00:18

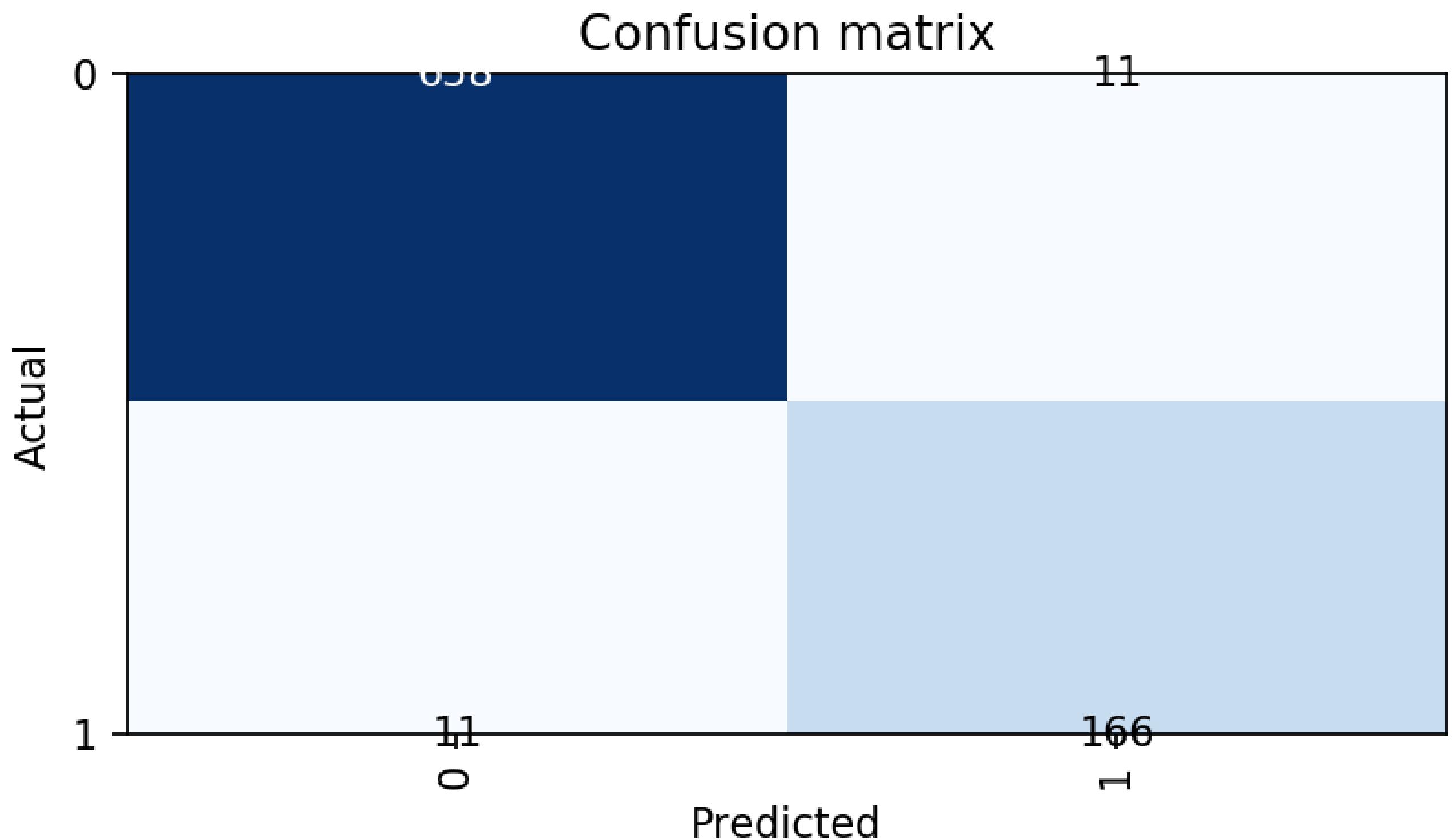


```
In [41]: interp = ClassificationInterpretation.from_learner(learner)
losses,idxs = interp.top_losses()
len(data.valid_ds)==len(losses)==len(idxs)
# plot
interp.plot_top_losses(9, figsize=(16,16))
```

**prediction/actual/loss/probability**

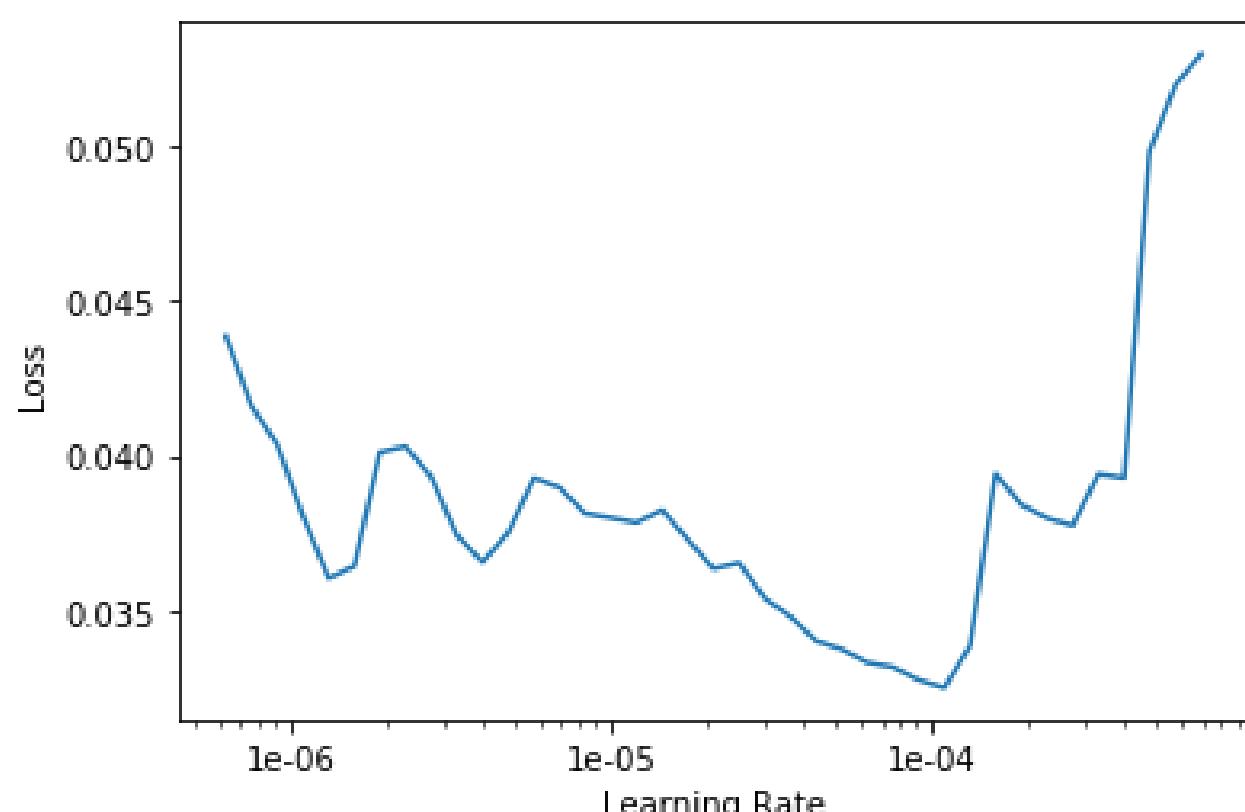


```
In [42]: interp.plot_confusion_matrix(figsize=(5,5), dpi=160)
```



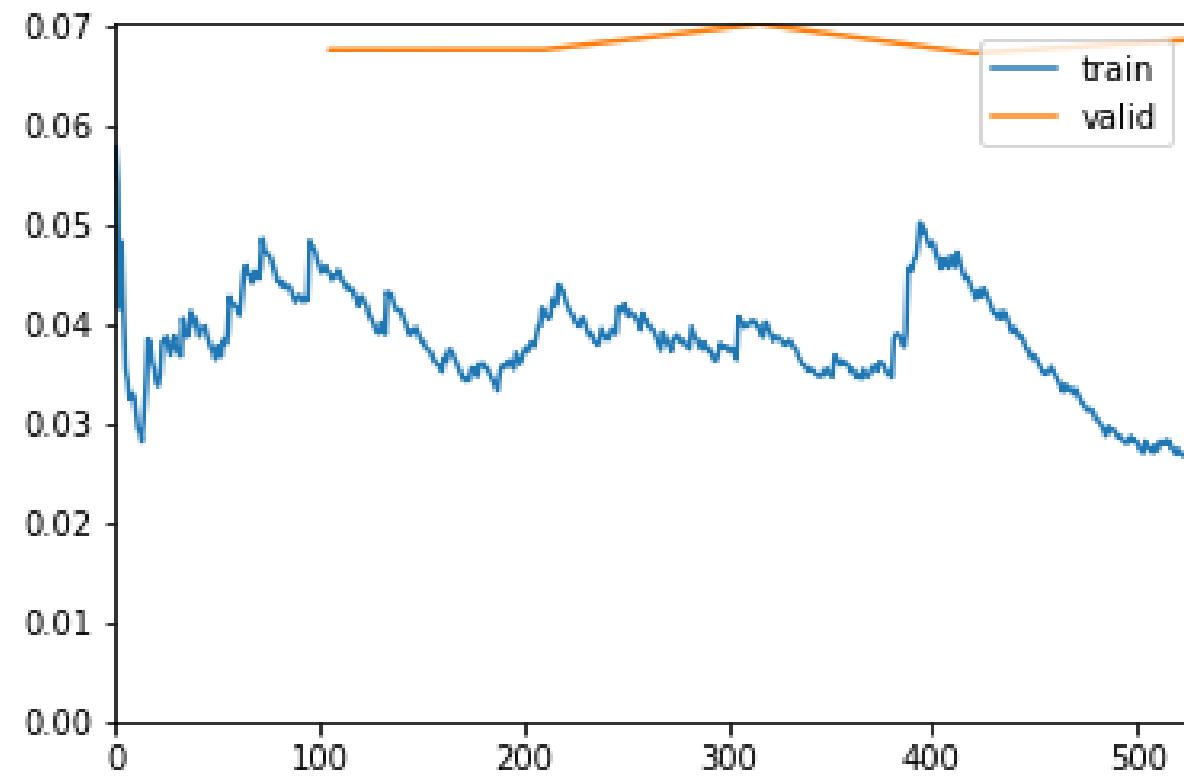
```
In [43]: learner.lr_find()  
learner.recorder.plot()
```

LR Finder is complete, type {learner\_name}.recorder.plot() to see the graph.



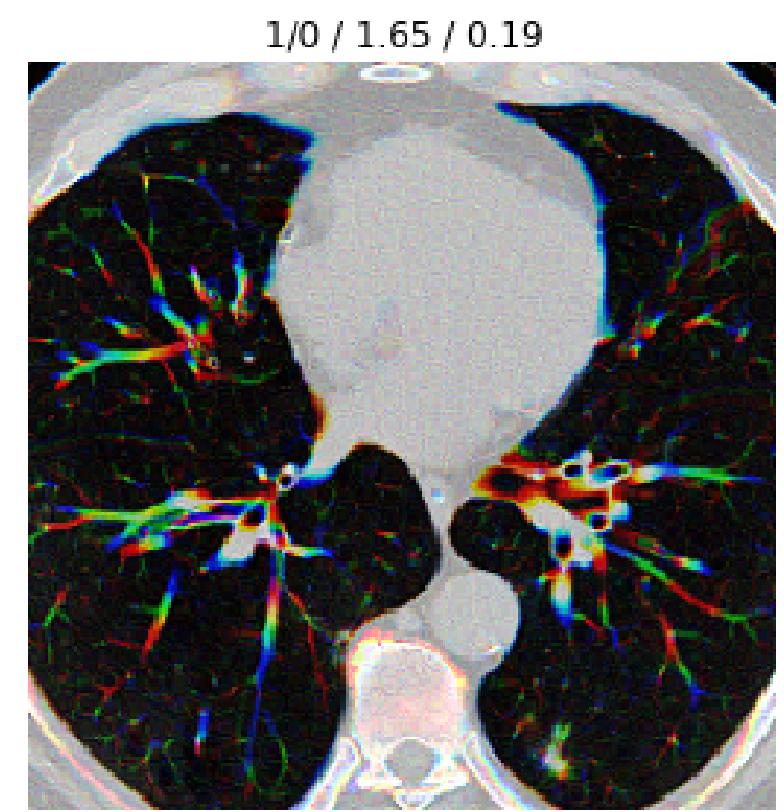
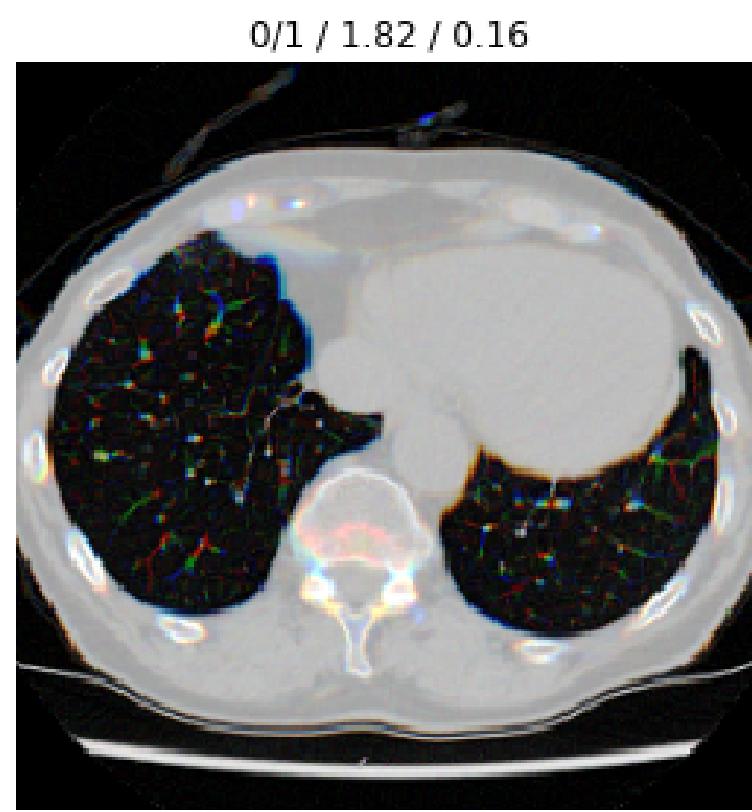
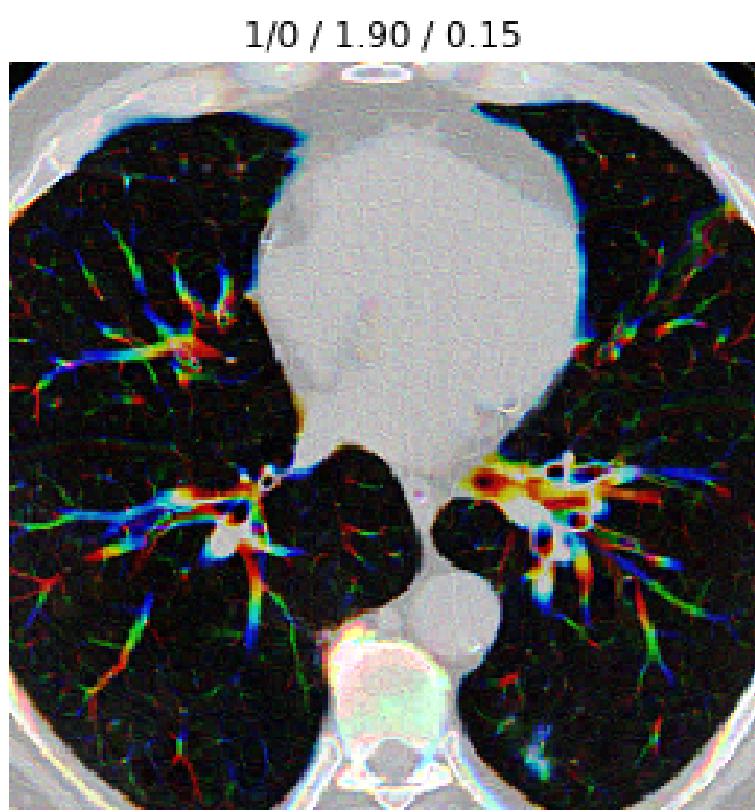
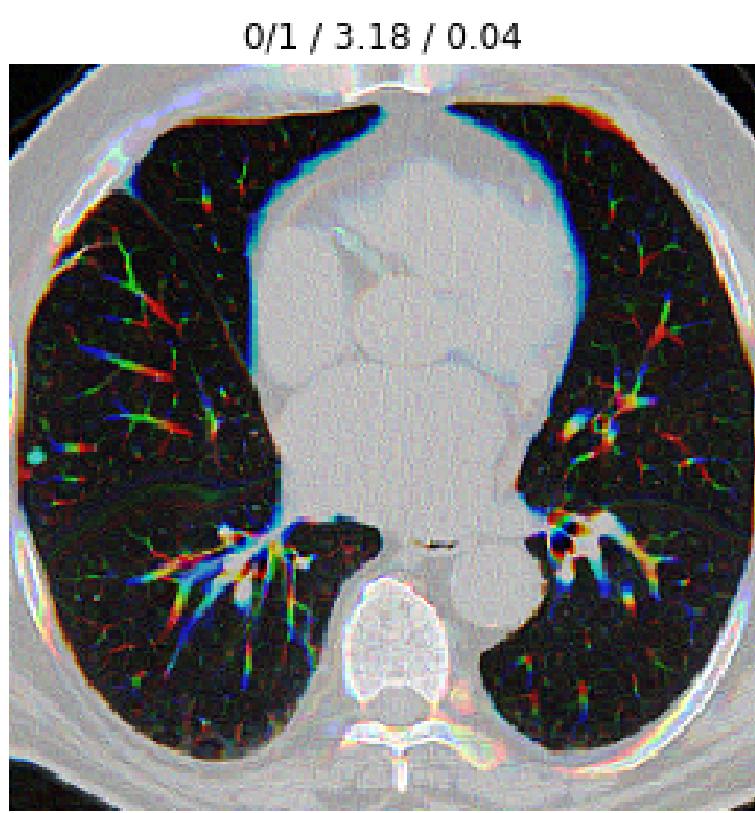
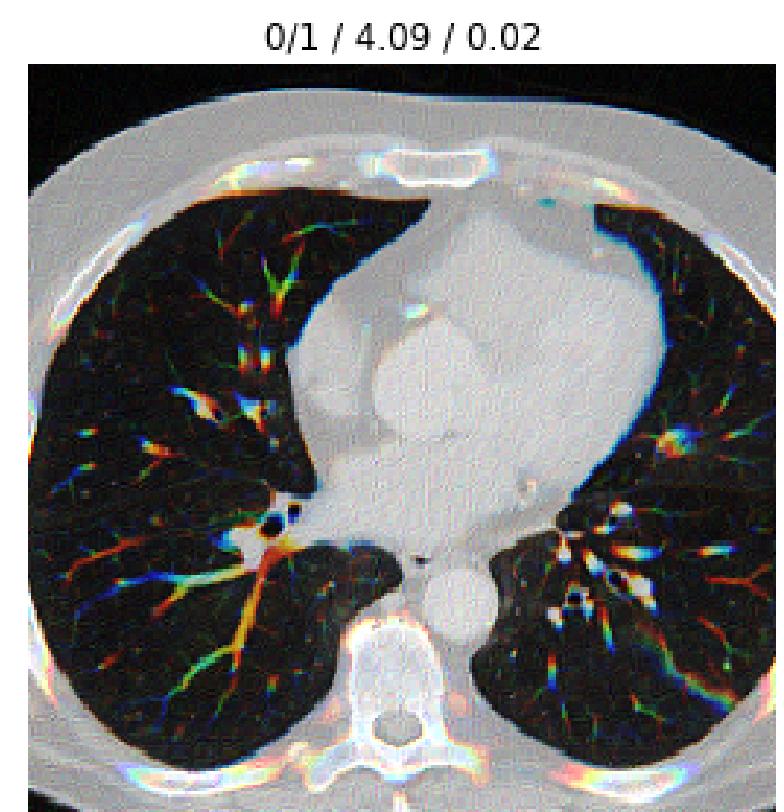
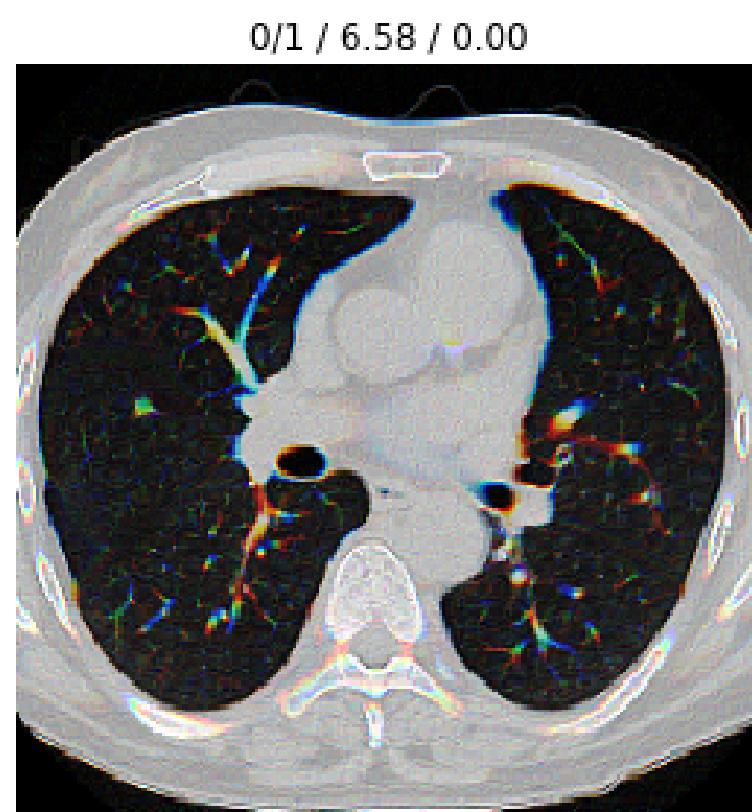
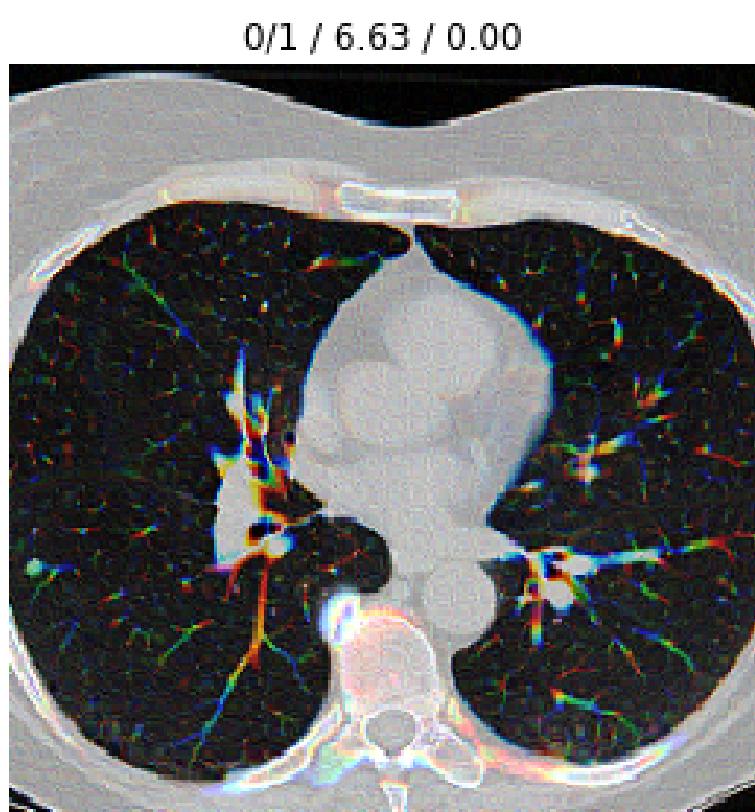
```
In [44]: learner.unfreeze()  
learner.fit_one_cycle(5, max_lr=1e-5)
```

epoch	train_loss	valid_loss	error_rate	auroc	time
0	0.045264	0.067641	0.018913	0.991656	00:17
1	0.041613	0.067668	0.023641	0.991711	00:18
2	0.039806	0.070195	0.018913	0.990972	00:18
3	0.043492	0.067287	0.020095	0.989942	00:18
4	0.026908	0.068629	0.020095	0.990178	00:18

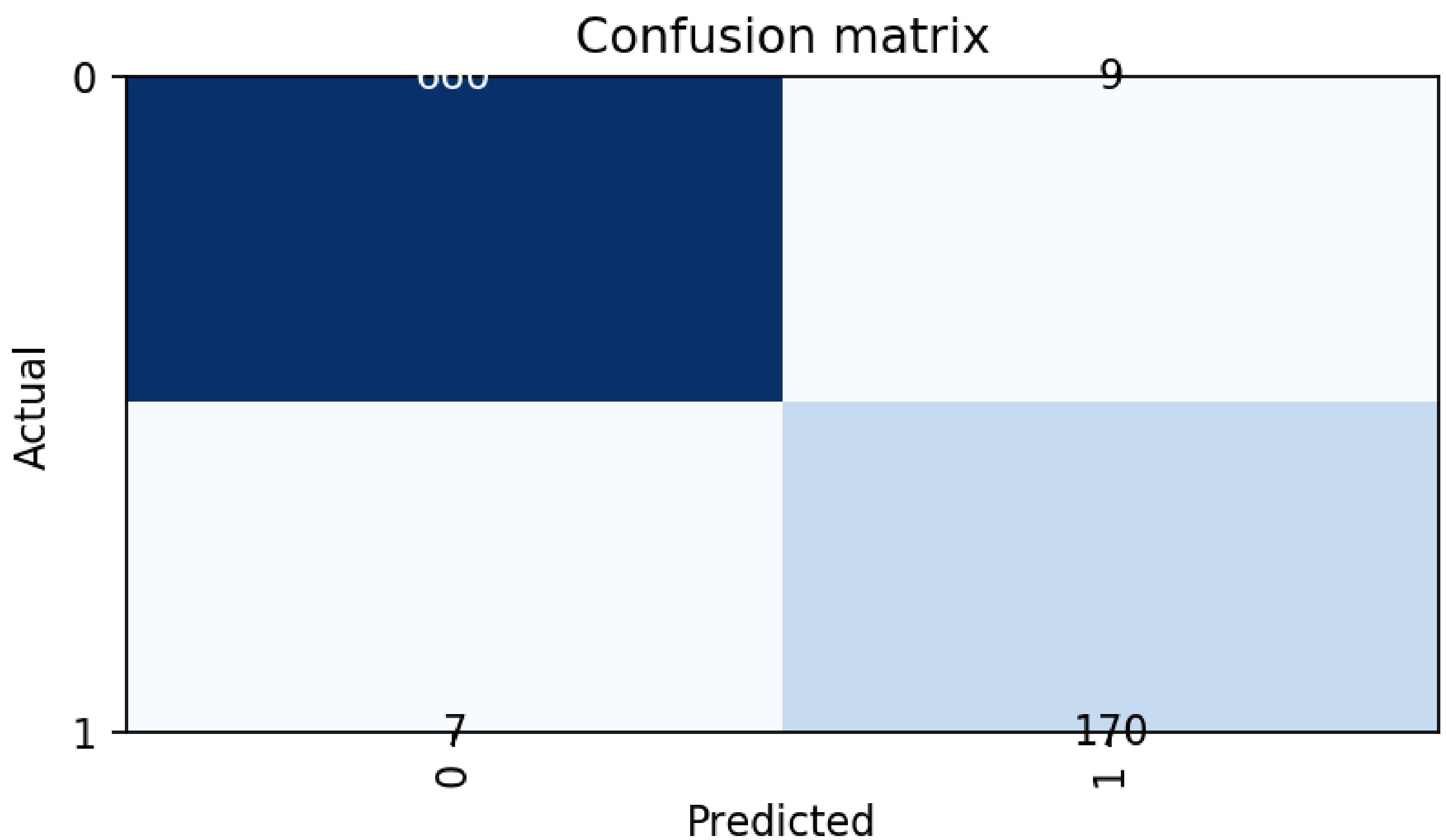


```
In [51]: interp = ClassificationInterpretation.from_learner(learner)
losses,idxs = interp.top_losses()
len(data.valid_ds)==len(losses)==len(idxs)
# plot
interp.plot_top_losses(9, figsize=(16,16))
```

**prediction/actual/loss/probability**

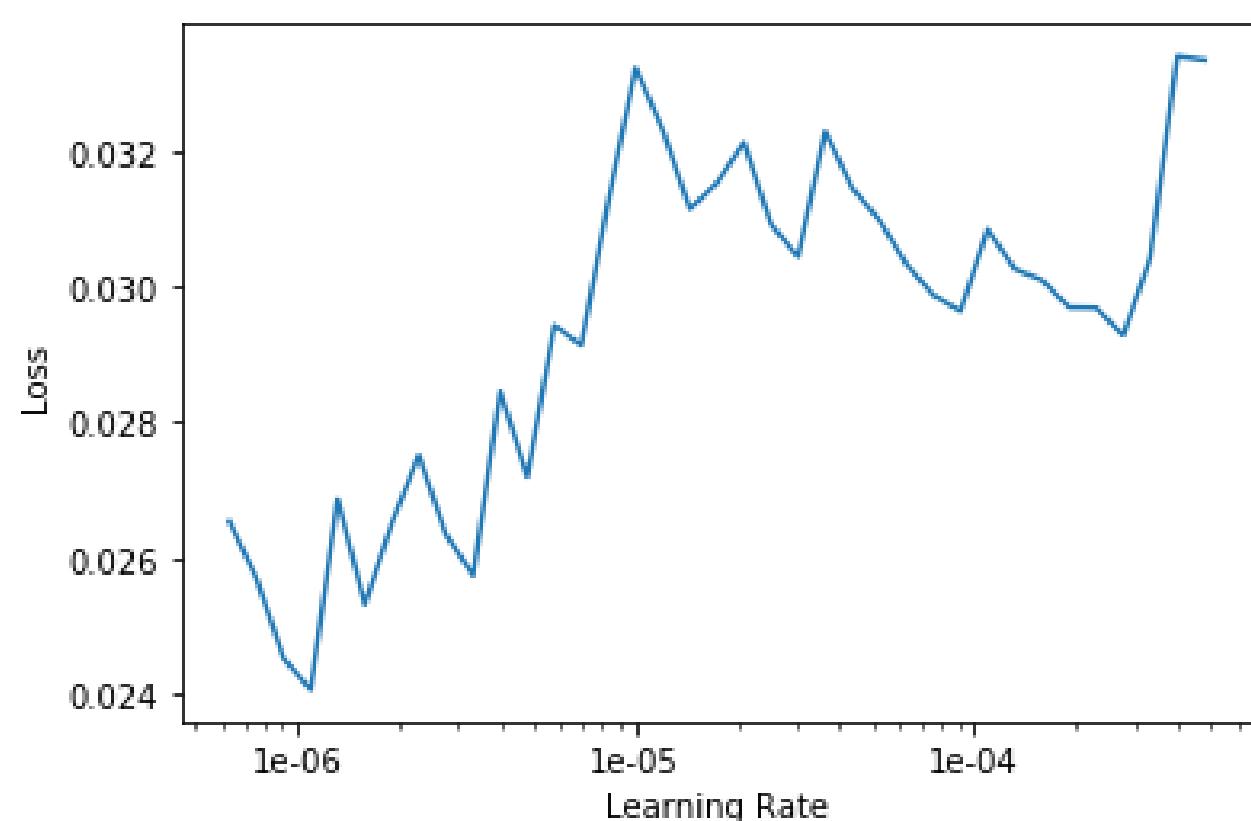


```
In [52]: interp.plot_confusion_matrix(figsize=(5,5), dpi=160)
```



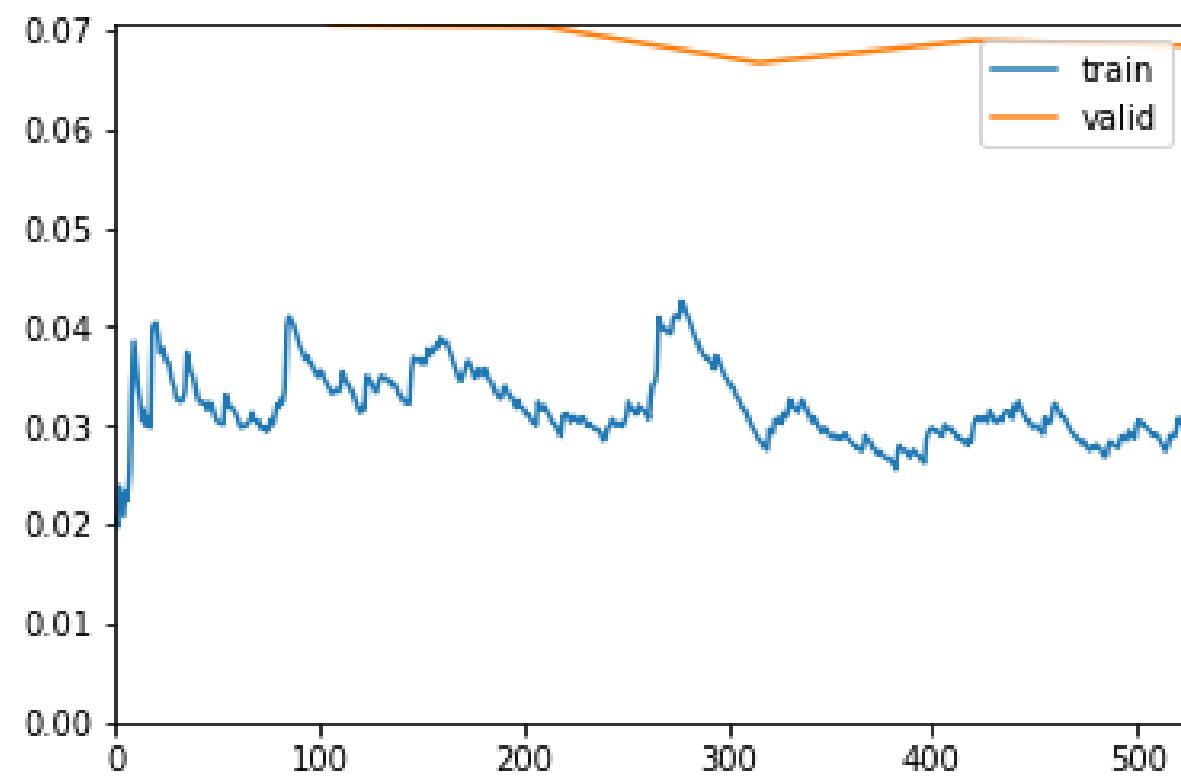
```
In [49]: learner.lr_find()  
learner.recorder.plot()
```

LR Finder is complete, type {learner\_name}.recorder.plot() to see the graph.



```
In [50]: learner.unfreeze()  
learner.fit_one_cycle(5, max_lr=1e-6)
```

epoch	train_loss	valid_loss	error_rate	auroc	time
0	0.034001	0.070497	0.023641	0.990153	00:18
1	0.031622	0.070329	0.020095	0.990288	00:18
2	0.029037	0.066694	0.021277	0.990111	00:18
3	0.028435	0.068945	0.024823	0.990626	00:18
4	0.029930	0.068407	0.018913	0.990639	00:17



```
In [ ]: # # get most confused filenames  
# losses,idxs = interp.top_losses(10)  
# for p in data.valid_ds.x.items[idxs]:  
#     print(p)
```

```
In [ ]: learner.save('stage-2-13epc')
```

```
In [53]: learner.export('vgg16-25epc-rescale-crop256-unbalanced')
```