Image segmentation with CAMVID

```
%reload ext autoreload
%autoreload 2
%matplotlib inline
from fastai.vision import *
from fastai.callbacks.hooks import *
from fastai.utils.mem import *
path = untar_data(URLs.CAMVID)
    Downloading <a href="https://s3.amazonaws.com/fast-ai-imagelocal/camvid">https://s3.amazonaws.com/fast-ai-imagelocal/camvid</a>
path.ls()
     [PosixPath('/root/.fastai/data/camvid/images'),
      PosixPath('/root/.fastai/data/camvid/valid.txt'),
      PosixPath('/root/.fastai/data/camvid/codes.txt'),
      PosixPath('/root/.fastai/data/camvid/labels')]
path_lbl = path/'labels'
path_img = path/'images'
fnames = get_image_files(path_img)
fnames[:3]
     [PosixPath('/root/.fastai/data/camvid/images/0016E5_08159.png'),
      PosixPath('/root/.fastai/data/camvid/images/0016E5_05010.png'),
      PosixPath('/root/.fastai/data/camvid/images/0016E5_08580.png')]
lbl_names = get_image_files(path_lbl)
lbl names[:3]
     [PosixPath('/root/.fastai/data/camvid/labels/0016E5 00720 P.png'),
      PosixPath('/root/.fastai/data/camvid/labels/0016E5 07560 P.png'),
      PosixPath('/root/.fastai/data/camvid/labels/0001TP 008760 P.png')]
img f = fnames[0]
img = open_image(img_f)
img.show(figsize=(5,5))
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```



convert image file names to equivalent label file names
get_y_fn = lambda x: path_lbl/f'{x.stem}_P{x.suffix}'

mask = open_mask(get_y_fn(img_f))
mask.show(figsize=(5,5), alpha=1)



src_size = np.array(mask.shape[1:])
src_size,mask.data

codes = np.loadtxt(path/'codes.txt', dtype=str); codes

```
size = src_size//2
free = gpu_mem_get_free_no_cache()
```

```
# The max size of bs depends on the available GPU KAM

if free > 8200: bs=8

else: bs=4

print(f"using bs={bs}, have {free}MB of GPU RAM free")

\[ \text{\text{cannwe}} \]

using bs=8, have 16270MB of GPU RAM free

\[ \text{src} = (SegmentationItemList.from_folder(path_img) \\
\text{.split_by_fname_file('../valid.txt')} \\
\text{.label_from_func(get_y_fn, classes=codes))}

\[ \text{data} = (\text{src.transform(get_transforms(), size=size, tfm_y=True)} \\
\text{.databunch(bs=bs)} \\
\text{.normalize(imagenet_stats))}

\[ \text{data.show_batch(2, figsize=(10,7))} \]
```











data.show_batch(2, figsize=(10,7), ds_type=DatasetType.Valid)

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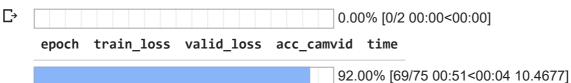




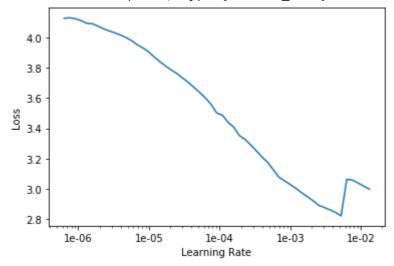




Modelling



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



lr=3e-3
learn.fit_one_cycle(10, slice(lr), pct_start=0.9)

₽	epoch	train_loss	valid_loss	acc_camvid	time
	0	1.196258	0.769203	0.827455	00:54
	1	0.818703	0.693821	0.826269	00:51
	2	0.722321	0.534580	0.856703	00:51
	3	0.634283	0.617080	0.850598	00:51
	4	0.629293	0.527589	0.858258	00:51
	5	0.582632	0.510770	0.871061	00:51
	6	0.636059	0.532386	0.857269	00:51
	7	0.674969	0.444165	0.872903	00:51
	8	0.575881	0.645875	0.839219	00:51
	9	0.489502	0.363996	0.895137	00:51

learn.save('stage-1')

learn.load('stage-1');

learn.show_results(rows=3, figsize=(8,9))

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Ground truth/Predictions













learn.unfreeze()

lrs = slice(lr/400,lr/4)

learn.fit_one_cycle(12, lrs, pct_start=0.8)

₽

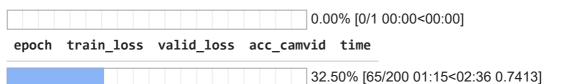
lr_find(learn)

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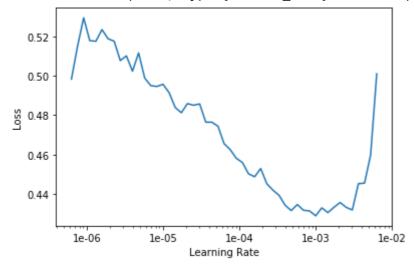
learn.recorder.plot()

	epoch	train_loss	valid_loss	acc_camvid	time		
	0	0.418384	0.350688	0.897273	00:53		
	1	0.406504	0.350110	0.895977	00:52		
	2	0.394534	0.334460	0.906691	00:52		
	3	0.381445	0.317356	0.907634	00:52		
	4	0.366837	0.315390	0.911705	00:53		
	5	0.353838	0.299530	0.918435	00:53		
	6	0.347749	0.296001	0.914528	00:53		
	7	0.330103	0.309215	0.910542	00:53		
	8	0.322873	0.292945	0.912567	00:53		
	9	0.314060	0.287867	0.922294	00:53		
	10	0.286326	0.286936	0.920011	00:53		
	11	0.262034	0.264440	0.925488	00:53		
	<pre>learn.save('stage-2');</pre>						
learn	.destro	y()					
size	= src_s	ize					
<pre>free = gpu_mem_get_free_no_cache() # the max size of bs depends on the available GPU RAM if free > 8200: bs=3 else:</pre>							
this Learner object self-destroyed - it still exists, but no longer usable using bs=3, have 15565MB of GPU RAM free							
<pre>data = (src.transform(get_transforms(), size=size, tfm_y=True)</pre>							
<pre>learn = unet_learner(data, models.resnet34, metrics=metrics, wd=wd)</pre>							
<pre>learn.load('stage-2');</pre>							

https://colab.research.google.com/drive/1wAljbljbQ2uAa-KrAEB8wy3dYWX2jSSB?authuser=1#scrollTo=1xtTCejVQzME&printMode=true



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



lr=1e-3
learn.fit_one_cycle(10, slice(lr), pct_start=0.8)

₽	epoch	train_loss	valid_loss	acc_camvid	time
	0	0.373758	0.313761	0.915802	03:38
	1	0.336631	0.314733	0.914408	03:35
	2	0.333394	0.287228	0.921204	03:35
	3	0.343204	0.303790	0.913903	03:35
	4	0.341679	0.291435	0.922916	03:35
	5	0.344433	0.350833	0.911346	03:35
	6	0.328087	0.298650	0.916003	03:35
	7	0.318760	0.306280	0.920618	03:35
	8	0.278263	0.275225	0.932161	03:35
	9	0.240248	0.240807	0.934117	03:35

learn.save('stage-1-big')

learn.load('stage-1-big');

learn.unfreeze()

lrs = slice(1e-6, lr/10)

learn.fit one cvcle(10. lrs)

₽	epoch	train_loss	valid_loss	acc_camvid	time
	0	0.224188	0.246791	0.934003	03:42
	1	0.213088	0.256830	0.932385	03:41
	2	0.218712	0.234583	0.936125	03:41
	3	0.218922	0.249010	0.932935	03:42
	4	0.219142	0.230402	0.937171	03:41
	5	0.204338	0.242721	0.935733	03:42
	6	0.196909	0.241375	0.936009	03:42
	7	0.199717	0.235098	0.937353	03:42
	8	0.191846	0.234249	0.937417	03:42
	9	0.193807	0.240966	0.936277	03:42

learn.save('stage-2-big')

learn.load('stage-2-big');

learn.show_results(rows=3, figsize=(10,10))

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Ground truth/Predictions











