```
In [1]: import sys
    sys.path.append('../deepsurv')
    import deep_surv

from deepsurv_logger import DeepSurvLogger, TensorboardLogger
    import utils
    import viz

import numpy as np
    import pandas as pd

import lasagne
    import matplotlib
    import matplotlib.pyplot as plt
%matplotlib inline
```

/Users/timfingerhut/anaconda3/lib/python3.6/site-packages/h5py/__i
nit__.py:36: FutureWarning: Conversion of the second argument of i
ssubdtype from `float` to `np.floating` is deprecated. In future,
it will be treated as `np.float64 == np.dtype(float).type`.
from . conv import register converters as register converters

```
In [2]: train_dataset_fp = './2006_final.csv'
    train_df = pd.read_csv(train_dataset_fp)
    train_df.head()
```

Out[2]:

	fail_1	dur	country1	country2	country4	country5	country6	country7	country8	count
0	1	30	1	0	0	0	0	0	0	
1	1	1	1	0	0	0	0	0	0	
2	1	60	1	0	0	0	0	0	0	
3	1	60	1	0	0	0	0	0	0	
4	1	45	1	0	0	0	0	0	0	

 $5 \text{ rows} \times 67 \text{ columns}$

```
In [3]: # event col is the header in the df that represents the 'Event / St
        atus' indicator
        # time col is the header in the df that represents the event time
        def dataframe to deepsurv ds(df, event col = 'fail 1', time col = '
        dur'):
            # Extract the event and time columns as numpy arrays
            e = df[event col].values.astype(np.int32)
            t = df[time col].values.astype(np.float32)
            # Extract the patient's covariates as a numpy array
            x df = df.drop([event col, time col], axis = 1)
            x = x df.values.astype(np.float32)
            # Return the deep surv dataframe
            return {
                x' : x'
                'e' : e,
                't' : t
            }
        # If the headers of the csv change, you can replace the values of
        # 'event col' and 'time col' with the names of the new headers
        # You can also use this function on your training dataset, validati
        on dataset, and testing dataset
        train data = dataframe to deepsurv ds(train df, event col = 'fail 1
         ', time_col= 'dur')
In [4]: hyperparams = {
            'L2 reg': 10.0,
            'batch norm': True,
            'dropout': 0.4,
             'hidden_layers_sizes': [100, 100],
             'learning rate': 2e-2,
             'lr_decay': 0.001,
             'momentum': 0.9,
            'n in': train data['x'].shape[1],
```

'standardize': True

}

```
model = deep surv.DeepSurv(**hyperparams)
# DeepSurv can now leverage TensorBoard to monitor training and val
idation
# This section of code is optional. If you don't want to use the te
nsorboard logger
# Uncomment the below line, and comment out the other three lines:
# logger = None
experiment name = 'test experiment tim'
logdir = './logs/tensorboard/'
logger = TensorboardLogger(experiment name, logdir=logdir)
# Now we train the model
update fn=lasagne.updates.amsgrad # The type of optimizer to use. \
                                      # Check out http://lasa
gne.readthedocs.io/en/latest/modules/updates.html \
                                      # for other optimizers
to use
n = pochs = 2000
# If you have validation data, you can add it as the second paramet
er to the function
metrics = model.train(train_data, n_epochs=n_epochs, logger=logger,
update fn=update fn)
2019-07-11 18:25:10,120 - Training step 0/2000
- loss: 28.5944 - ci: 0.5118
2019-07-11 18:26:47,626 - Training step 250/2000
- loss: 6.4920 - ci: 0.8007
2019-07-11 18:28:15,479 - Training step 500/2000
- loss: 6.4786 - ci: 0.8359
2019-07-11 18:29:43,128 - Training step 750/2000 | *******
- loss: 6.2922 - ci: 0.8726
2019-07-11 18:31:14,021 - Training step 1000/2000 | *********
- loss: 6.2526 - ci: 0.8836
- loss: 6.2010 - ci: 0.8931
- loss: 6.1536 - ci: 0.8978
- loss: 6.1427 - ci: 0.9004
2019-07-11 18:38:08,096 - Finished Training with 2000 iterations i
```

n 778.46s

In [5]: # Create an instance of DeepSurv using the hyperparams defined abov

```
In [6]: # Print the final metrics
    print('Train C-Index:', metrics['c-index'][-1])
    # print('Valid C-Index: ',metrics['valid_c-index'][-1])

# Plot the training / validation curves
    viz.plot_log(metrics)
```

Train C-Index: (1999, 0.9044605031983957)



