## Model

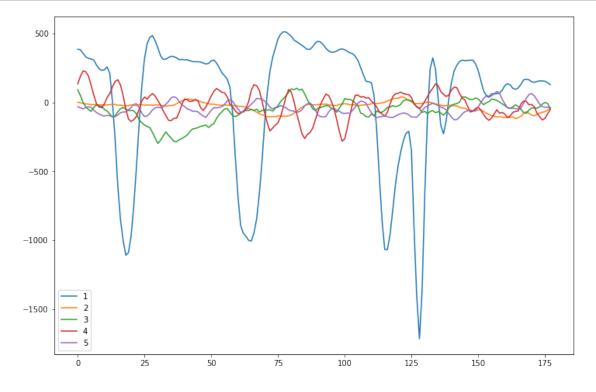
## February 1, 2021

[1]: import tensorflow as tf

```
import numpy as np
     from keras.models import Sequential
     from keras.utils import np_utils
     from keras.layers import Dense,Activation,LSTM,Dropout,AveragePooling3D
     import pandas as pd
     import matplotlib.pyplot as plt
[2]: df= pd.read_csv('data.csv')
     df.head()
[2]:
        Unnamed: 0
                           Х2
                                ХЗ
                                     Х4
                                           Х5
                                                Х6
                                                          X8
                                                                Х9
                                                                       X170
                                                                             X171
                     Х1
                                                     Х7
                               229
     0 X21.V1.791
                    135
                          190
                                    223
                                         192
                                               125
                                                     55
                                                          -9
                                                              -33
                                                                        -17
                                                                              -15
     1
       X15.V1.924
                    386
                          382
                               356
                                    331
                                         320
                                               315
                                                    307
                                                         272
                                                              244
                                                                        164
                                                                               150
     2
           X8.V1.1 -32
                         -39
                               -47
                                    -37
                                          -32
                                               -36
                                                    -57
                                                         -73
                                                              -85
                                                                         57
                                                                               64
     3
         X16.V1.60 -105 -101
                               -96
                                    -92
                                         -89
                                               -95 -102 -100
                                                              -87
                                                                        -82
                                                                              -81
                              -98 -102
                                         -78
         X20.V1.54
                      -9
                         -65
                                               -48
                                                    -16
                                                           0
                                                              -21
                                                                          4
                                                                                2
        X172
              X173
                    X174
                           X175
                                 X176
                                       X177
                                              X178
                                                    У
         -31
               -77
                    -103
                           -127
                                 -116
                                        -83
     0
                                               -51
                                                    4
         146
     1
               152
                      157
                            156
                                  154
                                        143
                                               129
                                                   1
     2
          48
                19
                      -12
                            -30
                                  -35
                                        -35
                                               -36 5
     3
         -80
               -77
                      -85
                            -77
                                               -65 5
                                  -72
                                        -69
         -12
               -32
                      -41
                            -65
                                  -83
                                        -89
                                               -73 5
     [5 rows x 180 columns]
[3]: x=df.values
     x=x[:,1:-1]
     x = np.asarray(x).astype(np.float32)
[4]: from sklearn.model_selection import train_test_split
     ##taking 20% for test
     y=np.array(df['y'])
     y=np_utils.to_categorical(y)
     print(y.shape)
```

```
(11500, 6)
(11500, 178)
```

```
[5]: ##corresponding to the 5 datasets ,plotting them all alongsize
plt.figure(figsize=(12,8))
plt.plot(x[1,:],label='1')
plt.plot(x[7,:],label='2')
plt.plot(x[12,:],label='3')
plt.plot(x[0,:],label='4')
plt.plot(x[2,:],label='5')
plt.legend()
plt.show()
```



```
[6]: ##makingg a neural network
model=Sequential()

model.add(Dense(256,input_shape=(45,)))
model.add(Activation('relu'))
model.add(Dense(128))
model.add(Activation('relu'))
```

```
model.add(Dense(128))
model.add(Activation('relu'))
model.add(Dense(64))
model.add(Activation('relu'))
model.add(Dense(32))
model.add(Activation('relu'))
model.add(Dense(32))
model.add(Activation('relu'))
model.add(Dense(32))
model.add(Activation('relu'))
model.add(Dense(16))
model.add(Activation('relu'))
model.add(Dense(16))
model.add(Activation('relu'))
model.add(Dense(8))
model.add(Activation('relu'))
model.add(Dense(5))
model.add(Activation('softmax'))
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 256)	11776
activation (Activation)	(None, 256)	0
dense_1 (Dense)	(None, 128)	32896
activation_1 (Activation)	(None, 128)	0
dense_2 (Dense)	(None, 128)	16512
activation_2 (Activation)	(None, 128)	0
dense_3 (Dense)	(None, 64)	8256
activation_3 (Activation)	(None, 64)	0
dense_4 (Dense)	(None, 32)	2080
activation_4 (Activation)	(None, 32)	0
dense_5 (Dense)	(None, 32)	1056
activation_5 (Activation)	(None, 32)	0

```
dense_6 (Dense)
                            (None, 32)
                                                    1056
    activation_6 (Activation) (None, 32)
   dense_7 (Dense)
                             (None, 16)
   activation_7 (Activation) (None, 16)
   dense_8 (Dense)
                             (None, 16)
                                                    272
   activation_8 (Activation) (None, 16)
   dense_9 (Dense)
                            (None, 8)
                                                    136
   activation_9 (Activation) (None, 8)
   dense_10 (Dense)
                             (None, 5)
    activation_10 (Activation) (None, 5)
    ______
   Total params: 74,613
   Trainable params: 74,613
   Non-trainable params: 0
    ______
[7]: model.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])
[9]: ##batch size : 15 ,epochs : 50
    hist = model.fit(((x_train[:,::4]-x_train.mean())/x_train.std()),y_train[:,1:
     \rightarrow],15,50,verbose=2,validation_data=((x_test[:,::4]-x_test.mean())/x_test.
     →std(),y_test[:,1:]),shuffle=False)
   614/614 - 3s - loss: 0.4146 - accuracy: 0.4079 - val_loss: 0.3813 -
   val accuracy: 0.4770
   Epoch 2/50
   614/614 - 2s - loss: 0.3641 - accuracy: 0.4917 - val_loss: 0.3635 -
   val_accuracy: 0.4848
   Epoch 3/50
   614/614 - 2s - loss: 0.3130 - accuracy: 0.5800 - val_loss: 0.2886 -
   val_accuracy: 0.6252
   Epoch 4/50
   614/614 - 2s - loss: 0.2488 - accuracy: 0.6786 - val_loss: 0.2592 -
   val_accuracy: 0.6678
   Epoch 5/50
   614/614 - 2s - loss: 0.2276 - accuracy: 0.7065 - val_loss: 0.2615 -
   val_accuracy: 0.6648
```

```
Epoch 6/50
614/614 - 2s - loss: 0.2146 - accuracy: 0.7211 - val_loss: 0.2604 -
val_accuracy: 0.6804
Epoch 7/50
614/614 - 2s - loss: 0.2054 - accuracy: 0.7354 - val_loss: 0.2694 -
val_accuracy: 0.6652
Epoch 8/50
614/614 - 2s - loss: 0.1909 - accuracy: 0.7566 - val_loss: 0.2708 -
val_accuracy: 0.6813
Epoch 9/50
614/614 - 2s - loss: 0.1872 - accuracy: 0.7708 - val_loss: 0.2675 -
val_accuracy: 0.6848
Epoch 10/50
614/614 - 2s - loss: 0.1692 - accuracy: 0.7924 - val_loss: 0.2989 -
val_accuracy: 0.6900
Epoch 11/50
614/614 - 2s - loss: 0.1569 - accuracy: 0.8142 - val_loss: 0.2925 -
val_accuracy: 0.6874
Epoch 12/50
614/614 - 2s - loss: 0.1539 - accuracy: 0.8218 - val_loss: 0.3183 -
val_accuracy: 0.6891
Epoch 13/50
614/614 - 2s - loss: 0.1471 - accuracy: 0.8352 - val_loss: 0.3278 -
val_accuracy: 0.6883
Epoch 14/50
614/614 - 2s - loss: 0.1295 - accuracy: 0.8579 - val_loss: 0.3381 -
val_accuracy: 0.6857
Epoch 15/50
614/614 - 2s - loss: 0.1251 - accuracy: 0.8700 - val_loss: 0.2993 -
val_accuracy: 0.6830
Epoch 16/50
614/614 - 2s - loss: 0.1149 - accuracy: 0.8830 - val_loss: 0.3701 -
val_accuracy: 0.6922
Epoch 17/50
614/614 - 2s - loss: 0.1074 - accuracy: 0.8899 - val loss: 0.3547 -
val_accuracy: 0.7009
Epoch 18/50
614/614 - 2s - loss: 0.0960 - accuracy: 0.9052 - val_loss: 0.4338 -
val_accuracy: 0.7009
Epoch 19/50
614/614 - 2s - loss: 0.0946 - accuracy: 0.9052 - val_loss: 0.3940 -
val_accuracy: 0.7017
Epoch 20/50
614/614 - 2s - loss: 0.0891 - accuracy: 0.9145 - val_loss: 0.3721 -
val_accuracy: 0.7061
Epoch 21/50
614/614 - 2s - loss: 0.0894 - accuracy: 0.9123 - val_loss: 0.4215 -
val_accuracy: 0.7035
```

```
Epoch 22/50
614/614 - 2s - loss: 0.0773 - accuracy: 0.9242 - val_loss: 0.4184 -
val_accuracy: 0.7165
Epoch 23/50
614/614 - 2s - loss: 0.0708 - accuracy: 0.9314 - val loss: 0.4375 -
val_accuracy: 0.7057
Epoch 24/50
614/614 - 2s - loss: 0.0703 - accuracy: 0.9323 - val_loss: 0.4530 -
val_accuracy: 0.7043
Epoch 25/50
614/614 - 2s - loss: 0.0621 - accuracy: 0.9423 - val_loss: 0.4876 -
val_accuracy: 0.7091
Epoch 26/50
614/614 - 2s - loss: 0.0604 - accuracy: 0.9450 - val_loss: 0.5129 -
val_accuracy: 0.7065
Epoch 27/50
614/614 - 2s - loss: 0.0602 - accuracy: 0.9451 - val_loss: 0.4366 -
val_accuracy: 0.7139
Epoch 28/50
614/614 - 2s - loss: 0.0541 - accuracy: 0.9533 - val_loss: 0.4895 -
val_accuracy: 0.7087
Epoch 29/50
614/614 - 2s - loss: 0.0497 - accuracy: 0.9537 - val_loss: 0.5161 -
val_accuracy: 0.6996
Epoch 30/50
614/614 - 2s - loss: 0.0572 - accuracy: 0.9472 - val_loss: 0.5198 -
val_accuracy: 0.7043
Epoch 31/50
614/614 - 2s - loss: 0.0456 - accuracy: 0.9587 - val_loss: 0.5436 -
val_accuracy: 0.7052
Epoch 32/50
614/614 - 2s - loss: 0.0476 - accuracy: 0.9579 - val_loss: 0.5465 -
val_accuracy: 0.7174
Epoch 33/50
614/614 - 2s - loss: 0.0485 - accuracy: 0.9567 - val loss: 0.4898 -
val_accuracy: 0.7026
Epoch 34/50
614/614 - 2s - loss: 0.0434 - accuracy: 0.9605 - val_loss: 0.5539 -
val_accuracy: 0.7026
Epoch 35/50
614/614 - 2s - loss: 0.0368 - accuracy: 0.9659 - val_loss: 0.5763 -
val_accuracy: 0.7252
Epoch 36/50
614/614 - 2s - loss: 0.0356 - accuracy: 0.9691 - val_loss: 0.5746 -
val_accuracy: 0.7117
Epoch 37/50
614/614 - 2s - loss: 0.0349 - accuracy: 0.9702 - val_loss: 0.5417 -
val_accuracy: 0.7235
```

```
Epoch 38/50
     614/614 - 2s - loss: 0.0398 - accuracy: 0.9641 - val_loss: 0.5281 -
     val_accuracy: 0.7191
     Epoch 39/50
     614/614 - 2s - loss: 0.0383 - accuracy: 0.9680 - val loss: 0.5602 -
     val_accuracy: 0.6996
     Epoch 40/50
     614/614 - 2s - loss: 0.0313 - accuracy: 0.9714 - val_loss: 0.6495 -
     val_accuracy: 0.6965
     Epoch 41/50
     614/614 - 2s - loss: 0.0358 - accuracy: 0.9689 - val_loss: 0.5344 -
     val_accuracy: 0.7048
     Epoch 42/50
     614/614 - 2s - loss: 0.0318 - accuracy: 0.9717 - val_loss: 0.5590 -
     val_accuracy: 0.7122
     Epoch 43/50
     614/614 - 2s - loss: 0.0262 - accuracy: 0.9760 - val_loss: 0.6060 -
     val_accuracy: 0.7143
     Epoch 44/50
     614/614 - 2s - loss: 0.0252 - accuracy: 0.9783 - val_loss: 0.6264 -
     val_accuracy: 0.7048
     Epoch 45/50
     614/614 - 2s - loss: 0.0268 - accuracy: 0.9780 - val_loss: 0.6572 -
     val_accuracy: 0.7030
     Epoch 46/50
     614/614 - 2s - loss: 0.0329 - accuracy: 0.9764 - val_loss: 0.5634 -
     val_accuracy: 0.7096
     Epoch 47/50
     614/614 - 2s - loss: 0.0267 - accuracy: 0.9801 - val_loss: 0.5555 -
     val_accuracy: 0.7196
     Epoch 48/50
     614/614 - 2s - loss: 0.0264 - accuracy: 0.9780 - val_loss: 0.5460 -
     val_accuracy: 0.7265
     Epoch 49/50
     614/614 - 2s - loss: 0.0272 - accuracy: 0.9768 - val loss: 0.5831 -
     val_accuracy: 0.7013
     Epoch 50/50
     614/614 - 2s - loss: 0.0343 - accuracy: 0.9714 - val_loss: 0.6016 -
     val_accuracy: 0.7152
[12]: ## now with using lstm
      model = Sequential()
      model.add(LSTM(56, input_shape=(45,1), return_sequences=True))
      model.add(Dropout(0.3))
      model.add(LSTM(56))
      model.add(Dropout(0.3))
      model.add(Dense(20))
```

```
model.add(Activation('tanh'))
     model.add(Dense(5))
     model.add(Activation('softmax'))
    model.summary()
    Model: "sequential_1"
    Layer (type)
                            Output Shape
                                                  Param #
    ______
    1stm (LSTM)
                            (None, 45, 56)
                                                  12992
    dropout (Dropout)
                            (None, 45, 56)
    lstm_1 (LSTM)
                            (None, 56)
                                                 25312
    dropout_1 (Dropout)
                           (None, 56)
    dense_11 (Dense)
                           (None, 20)
                                                 1140
    activation_11 (Activation) (None, 20)
    _____
    dense_12 (Dense)
                           (None, 5)
                                                 105
    activation_12 (Activation) (None, 5)
                                                 Ω
    _____
    Total params: 39,549
    Trainable params: 39,549
    Non-trainable params: 0
[13]: model.compile(loss='binary_crossentropy', optimizer='adam',_
     →metrics=['accuracy'])
[16]: hist2 = model.fit(((x_train[:,::4]-x_train.mean())/x_train.std()),y_train[:,1:
     →], validation_data=((x_test[:,::4]-x_test.mean())/x_test.std(),y_test[:,1:])
                  ,epochs = 50, batch_size=15,shuffle=False
    Epoch 1/50
                                        Traceback (most recent call last)
     <ipython-input-16-d0af242c9119> in <module>
```

```
---> 1 hist2 = model.fit(((x train[:,::4]-x train.mean())/x train.
  \rightarrowstd()),y_train[:,1:],validation_data=((x_test[:,::4]-x_test.mean())/x_test.
  2
                           ,epochs = 50, batch_size=15,shuffle=False
       3
       4
 ~/.local/lib/python3.8/site-packages/tensorflow/python/keras/engine/training.py
  →in fit(self, x, y, batch_size, epochs, verbose, callbacks, validation_split, validation_data, shuffle, class_weight, sample_weight, initial_epoch, validation_steps, validation_batch_size, validation_freq, validation_freq, validation_steps
  →max queue size, workers, use multiprocessing)
    1098
                           r=1):
    1099
                         callbacks.on_train_batch_begin(step)
 -> 1100
                         tmp_logs = self.train_function(iterator)
    1101
                         if data handler.should sync:
    1102
                           context.async wait()
 ~/.local/lib/python3.8/site-packages/tensorflow/python/eager/def function.py in

    call_(self, *args, **kwds)

              tracing_count = self.experimental_get_tracing_count()
     826
     827
              with trace.Trace(self._name) as tm:
 --> 828
                result = self._call(*args, **kwds)
     829
                compiler = "xla" if self._experimental_compile else "nonXla"
                new_tracing_count = self.experimental_get_tracing_count()
     830
 ~/.local/lib/python3.8/site-packages/tensorflow/python/eager/def function.py in

→ call(self, *args, **kwds)
     860
                # In this case we have not created variables on the first call. So _{
m I}
  -we can
                # run the first trace but we should fail if variables are created
     861
 --> 862
                results = self. stateful fn(*args, **kwds)
                if self._created_variables:
     863
     864
                  raise ValueError("Creating variables on a non-first call to a
  →function"
 ~/.local/lib/python3.8/site-packages/tensorflow/python/eager/function.py in_
  → call (self, *args, **kwargs)
    2939
              with self._lock:
    2940
                (graph_function,
 -> 2941
                 filtered flat args) = self. maybe_define_function(args, kwargs)
    2942
              return graph function. call flat(
    2943
                  filtered_flat_args, captured_inputs=graph_function.
  →captured inputs) # pylint: disable=protected-access
 ~/.local/lib/python3.8/site-packages/tensorflow/python/eager/function.py in
  → maybe_define_function(self, args, kwargs)
    3355
                         self.input signature is None and
```

```
3356
                       call_context_key in self._function_cache.missed):
-> 3357
                     return self._define_function_with_shape_relaxation(
   3358
                         args, kwargs, flat_args, filtered_flat_args,_
→cache key context)
   3359
~/.local/lib/python3.8/site-packages/tensorflow/python/eager/function.py in_
→_define_function_with_shape_relaxation(self, args, kwargs, flat_args, __
→filtered_flat_args, cache_key_context)
   3277
                   expand composites=True)
   3278
-> 3279
            graph function = self. create graph function(
   3280
                args, kwargs, override_flat_arg_shapes=relaxed_arg_shapes)
            self. function cache.arg relaxed[rank only cache key] = ____
   3281
→graph_function
~/.local/lib/python3.8/site-packages/tensorflow/python/eager/function.py in_
→ create graph function(self, args, kwargs, override flat arg shapes)
            arg_names = base_arg_names + missing_arg_names
   3194
            graph function = ConcreteFunction(
   3195
-> 3196
                func_graph_module.func_graph_from_py_func(
   3197
                     self. name,
   3198
                     self._python_function,
~/.local/lib/python3.8/site-packages/tensorflow/python/framework/func_graph.py_
→in func_graph_from_py_func(name, python_func, args, kwargs, signature, →func_graph, autograph, autograph_options, add_control_dependencies, arg_names ⊔
→op_return_value, collections, capture_by_value, override_flat_arg_shapes)
                _, original_func = tf_decorator.unwrap(python_func)
    988
    989
--> 990
               func outputs = python func(*func args, **func kwargs)
    991
    992
               # invariant: `func outputs` contains only Tensors,
→CompositeTensors,
~/.local/lib/python3.8/site-packages/tensorflow/python/eager/def_function.py_in_
→wrapped_fn(*args, **kwds)
    632
                     xla context.Exit()
    633
                else:
--> 634
                   out = weak_wrapped_fn().__wrapped__(*args, **kwds)
    635
                return out
    636
~/.local/lib/python3.8/site-packages/tensorflow/python/framework/func graph.py_
→in wrapper(*args, **kwargs)
    975
                   except Exception as e: # pylint:disable=broad-except
    976
                     if hasattr(e, "ag error metadata"):
```

```
--> 977
                     raise e.ag_error_metadata.to_exception(e)
    978
                   else:
    979
                      raise
ValueError: in user code:
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/keras/
 →engine/training.py:805 train_function *
        return step_function(self, iterator)
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/keras/
 →engine/training.py:795 step_function **
        outputs = model.distribute_strategy.run(run_step, args=(data,))
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/distribut
 ⇒distribute_lib.py:1259 run
        return self._extended.call_for_each_replica(fn, args=args, kwargs=kwarg)
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/distribut
 →distribute_lib.py:2730 call_for_each_replica
        return self._call_for_each_replica(fn, args, kwargs)
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/distribut/
 →distribute_lib.py:3417 _call_for_each_replica
        return fn(*args, **kwargs)
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/keras/
 →engine/training.py:788 run_step **
        outputs = model.train step(data)
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/keras/
 →engine/training.py:754 train_step
        y_pred = self(x, training=True)
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/keras/
 →engine/base_layer.py:998 __call__
        input_spec.assert_input_compatibility(self.input_spec, inputs, self.nam)
    /home/sagnik/.local/lib/python3.8/site-packages/tensorflow/python/keras/
 →engine/input_spec.py:219 assert_input_compatibility
        raise ValueError('Input ' + str(input_index) + ' of layer ' +
    ValueError: Input 0 of layer sequential 1 is incompatible with the layer:
 →expected ndim=3, found ndim=2. Full shape received: (None, 45)
```

[]: