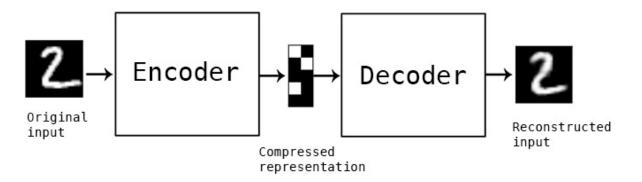
## **Encoder-Decoder Model for Sequence Prediction**

What are autoencoders?



"Autoencoding" is a data compression algorithm where the compression and decompression functions are 1) data-specific, 2) lossy, and 3) learned automatically from examples rather than engineered by a human. Additionally, in almost all contexts where the term "autoencoder" is used, the compression and decompression functions are implemented with neural networks.

Here, we will use an LSTM neural network to learn the encoding of a five digit sequence.

```
In [1]: # load libraries
    import os
    import pandas as pd
    import numpy as np
    from numpy import array
    from numpy import array_equal
    from numpy import argmax
    import matplotlib.pyplot as plt
    from sklearn.preprocessing import MinMaxScaler
    from keras.utils import to_categorical
    from keras.models import Model
    from keras.layers import Input, LSTM, Dense, Dropout
    from keras.callbacks import ModelCheckpoint
```

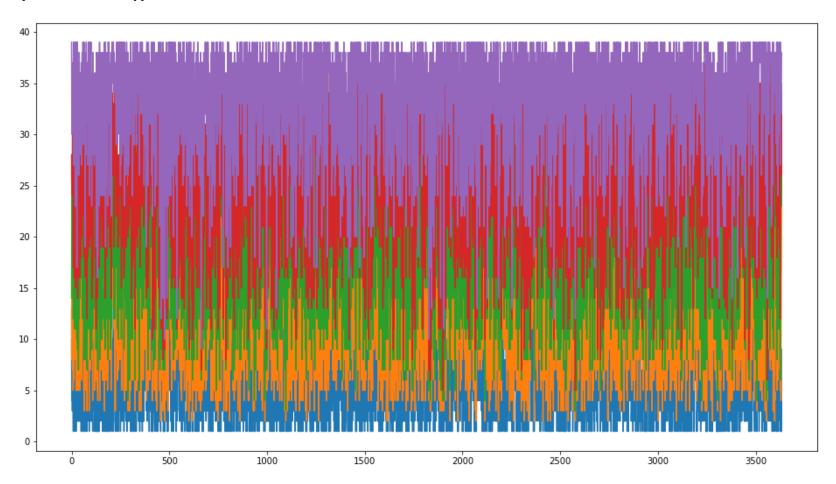
Using TensorFlow backend.

```
In [2]: | # set random seed
        seed = 777
        np.random.seed(seed)
In [3]: series = pd.read_csv('/Users/brent/Sandbox/ai/data/five_sequence.csv', engine='python', header=None)
        series.head(10)
Out[3]:
            0 1 2 3 4
        0 10 13 14 28 39
        1 4 8 21 25 33
        2 7 18 19 27 30
        3 12 17 25 26 30
        4 3 12 22 24 37
        5 6 7 20 33 34
        6 24 31 32 33 36
        7 17 21 30 33 36
        8 3 5 19 24 32
        9 1 6 12 17 34
```

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```
In [4]: # review dataset
    data = series.values
    print("Data shape: " + str(data.shape))
    print(data)
    plt.figure(figsize=(16,9))
    plt.plot(data)
    plt.show()
```

```
Data shape: (3633, 5)
[[10 13 14 28 39]
[ 4 8 21 25 33]
[ 7 18 19 27 30]
...
[ 4 16 26 29 32]
[ 1 17 23 26 35]
[ 9 24 31 34 38]]
```



```
In [5]: # create a base list of the series data
mylist = []
    for i in range(len(data)):
        temp = data[i]
        merge = temp.tolist()
        mylist.extend(merge)
    print(len(mylist))
    print(mylist)
```

18165 [10, 13, 14, 28, 39, 4, 8, 21, 25, 33, 7, 18, 19, 27, 30, 12, 17, 25, 26, 30, 3, 12, 22, 24, 37, 6, 7, 20, 33, 34, 24, 31, 32, 33, 36, 17, 21, 30, 33, 36, 3, 5, 19, 24, 32, 1, 6, 12, 17, 34, 6, 24, 30, 37, 39, 3, 6, 21, 22, 27, 5, 9, 15, 18, 29, 6, 11, 25, 33, 36, 8, 10, 15, 16, 30, 1, 8, 11, 20, 29, 5, 17, 20, 30, 39, 2, 24, 29, 36, 39, 18, 20, 21, 22, 23, 7, 10, 12, 15, 23, 7, 9, 11, 31, 39, 6, 17, 20, 27, 32, 9, 10, 21, 25, 27, 1, 5, 14, 31, 34, 10, 28, 32, 35, 39, 8, 18, 27, 29, 32, 6, 18, 22, 26, 39, 4, 5, 10, 16, 17, 14, 29, 31, 35, 38, 2, 10, 28, 29, 38, 10, 11, 17, 21, 39, 4, 5, 9, 14, 28, 3, 11, 20, 23, 26, 5, 12, 26, 30, 34, 18, 22, 27, 29, 37, 2, 5, 7, 28, 33, 4, 10, 11, 21, 31, 2, 11, 21, 28, 37, 8, 9, 12, 16, 26, 4, 27, 31, 36, 38, 9, 21, 33, 34, 37, 9, 16, 22, 28, 35, 4, 5, 14, 16, 35, 2, 19, 21, 24, 36, 12, 21, 29, 30, 36, 2, 22, 24, 26, 30, 19, 21, 23, 27, 28, 12, 15, 16, 18, 36, 6, 15, 19, 30, 34, 1, 23, 26, 28, 33, 4, 14, 23, 24, 37, 3, 11, 16, 20, 27, 4, 6, 23, 24, 33, 11, 18, 27, 34, 35, 5, 22, 23, 34, 38, 8, 12, 35, 36, 37, 2, 8, 18, 22, 33, 1, 10, 12, 25, 35, 3, 22, 35, 38, 39, 4, 14, 24, 29, 34, 21, 22, 27, 30, 34, 8, 23, 26, 27, 29, 2, 6, 8, 26, 35, 3, 5, 10, 18, 39, 5, 10, 18, 21, 39, 13, 23, 31, 36, 38, 1, 3, 11, 34, 36, 10, 17, 18, 35, 38, 1, 9, 22, 32, 34, 9, 22, 29, 32, 34, 3, 7, 12, 16, 19, 2, 6, 8, 19, 36, 1, 9, 15, 19, 37, 7, 10, 13, 28, 37, 12, 14, 24, 25, 28, 3, 5, 13, 34, 39, 13, 17, 22, 31, 39, 7, 14, 16, 36, 37, 7, 24, 29, 33, 36, 1, 3, 7, 12, 37, 7, 10, 14, 20, 33, 7, 12, 30, 31, 34, 2, 9, 11, 18, 31, 9, 15, 26, 33, 36, 13, 14, 18, 26, 27, 5, 10, 27, 30, 32, 3, 8, 12, 31, 37, 4, 10, 19, 38, 39, 3, 17, 18, 28, 31, 4, 7, 11, 32, 39, 1, 3, 19, 30, 35, 5, 24, 26, 27, 34, 3, 8, 15, 19, 24, 19, 25, 30, 33, 39, 2, 28, 29, 31, 34, 11, 13, 15, 21, 27, 2, 7, 8, 32, 37, 7, 8, 13, 25, 27, 3, 9, 18, 20, 38, 7, 9, 10, 24, 39, 8, 18, 25, 30, 39, 9, 16, 18, 27, 32, 7, 9, 10, 31, 36, 2, 8, 9, 16, 23, 7, 8, 13, 14, 29, 4, 10, 12, 27, 31, 9, 11, 18, 31, 34, 3, 9, 26, 29, 31, 1, 4, 11, 16, 28, 8, 25, 27, 29, 38, 3, 17, 27, 29, 33, 18, 21, 25, 27, 35, 19, 26, 30, 33, 36, 5, 16, 22, 27, 31, 7, 10, 22, 31, 32, 3, 12, 13, 23, 35, 10, 15, 22, 26, 27, 1, 12, 28, 32, 34, 4, 5, 22, 32, 33, 14, 16, 23, 24, 28, 10, 16, 25, 35, 36, 1, 6, 14, 18, 29, 4, 10, 18, 25, 29, 23, 24, 25, 35, 36, 10, 11, 16, 21, 30, 1, 6, 9, 11, 35, 2, 13, 15, 18, 19, 4, 8, 11, 34, 38, 1, 9, 16, 24, 30, 6, 7, 28, 30, 33, 3, 21, 27, 31, 33, 1, 5, 9, 14, 16, 2, 10, 17, 29, 36, 1, 20, 22, 29, 34, 1, 4, 7, 17, 18, 12, 15, 19, 31, 35, 1, 10, 19, 27, 37, 8, 15, 16, 18, 38, 8, 14, 15, 19, 37, 1, 22, 26, 31, 32, 1, 9, 15, 17, 25, 1, 16, 18, 20, 23, 5, 9, 12, 19, 32, 26, 30, 36, 37, 39, 2, 4, 8, 28, 31, 10, 17, 27, 32, 35, 5, 8, 9, 36, 38, 2, 7, 11, 19, 33, 8, 12, 18, 21, 24, 16, 26, 34, 36, 37, 21, 22, 24, 28, 30, 4, 15, 20, 26, 36, 7, 16, 20, 22, 28, 8, 23, 29, 35, 37, 3, 17, 20, 31, 38, 2, 8, 10, 34, 38, 4, 7, 15, 22, 32, 13, 14, 15, 22, 24, 1, 6, 12, 19, 33, 5, 12, 32, 33, 34, 7, 15, 16, 22, 32, 2, 6, 27, 31, 33, 15, 20, 22, 31, 36, 7, 10, 24, 31, 38, 5, 6, 17, 19, 35, 10, 12, 22, 25, 27, 9, 13, 17, 19, 24, 1, 8, 28, 33, 36, 1, 3, 11, 36, 39, 5, 7, 23, 25, 29, 3, 21, 23, 25, 27, 8, 9, 23, 33, 38, 4, 6, 9, 14, 36, 6, 10, 19, 25, 39, 1, 9, 12, 18, 19, 22, 24, 30, 33, 38, 2, 6, 12, 26, 27, 2, 12, 25, 31, 38, 13, 16, 19, 24, 31, 5, 15, 16, 18, 37, 14, 15, 16, 24, 25, 7, 14, 23, 24, 34, 1, 11, 16, 26, 34, 10, 14, 18, 22, 32, 2, 5, 10, 21, 23, 1, 7, 22, 33, 38, 15, 17, 22, 31, 36, 5, 7, 8, 22, 39, 1, 14, 25, 30, 39, 7, 28, 31, 34, 38, 1, 16, 18, 26, 31, 2, 13, 32, 38, 39, 9, 22, 25, 27, 32, 3, 12, 27, 33, 37, 6, 11, 12, 19, 37, 21, 23, 27, 31, 35, 7, 8, 19, 20, 24, 14, 23, 27, 28, 31, 23, 28, 29, 33, 36, 17, 21, 24, 29, 33, 1, 8, 9, 19, 31, 13, 19, 21, 34, 38, 4, 5, 29, 30, 35, 14, 16, 24, 27, 32, 7, 10, 11, 14, 22, 4, 31, 33, 36, 37, 13, 20, 22, 30, 38, 3, 13, 17, 26, 39, 10, 29, 33, 34, 38, 1, 25, 26, 30, 35, 5, 23, 27, 33, 38, 7, 17, 29, 30, 36, 5, 16, 25, 31, 39, 1, 16, 19, 28, 39, 10, 19, 26, 32, 39, 10, 13, 20, 27, 34, 11, 20, 32, 33, 35, 10, 14, 15, 24, 34, 6, 11, 12, 19, 38, 14, 17, 33, 34, 37, 10, 11, 12, 28, 34, 8, 21, 30, 33, 35, 2, 3, 14, 34, 37, 2, 11, 15, 32, 35, 1, 3, 25, 26, 33, 11, 20, 23, 27, 34, 8, 16, 24, 30, 34, 5, 10, 17, 29, 34, 9, 18, 20, 22, 29, 10, 26, 31, 35, 38, 11, 21, 23, 27, 39, 1, 9, 21, 28, 32, 25, 26, 32, 33, 37, 4, 17, 27, 28, 35, 3, 6, 11, 28, 29, 4, 17, 20, 26, 37, 2, 13, 15, 27, 35, 7, 16, 17, 19, 39, 4, 22, 23, 26, 39, 3, 7, 10, 19, 39, 1, 7, 13, 28, 33, 6, 12, 17, 23, 39, 3, 12, 17, 20, 36, 6, 24, 25, 26, 31, 9, 21, 22, 28, 35, 10, 19, 21, 25, 30, 5, 12, 18, 26, 39, 5, 9, 19, 25, 30, 8, 10, 23, 36, 37, 2, 8, 13, 30, 38, 1, 14, 22, 24, 25, 2, 10, 27, 35, 39, 12, 14, 19, 24, 37, 4, 12, 26, 32, 39, 10, 12, 13, 24, 27, 3, 6, 10, 15, 30, 2, 5, 13, 32, 39, 5, 6, 20, 21, 35, 15, 16, 17, 20, 28,

return sh\_target

```
In [6]: # create the initial source set
        def create_initial_source(data, inlength):
            isource = list()
            for i in range(inlength):
                s = data[i]
                isource.append(s)
            return isource
In [7]: # create target sets and subsequent source sets
        def create targets(data, outlength):
            target, targ, tar = list(), list(), list()
            for j in range(int((len(data) - outlength) / outlength)):
                k = j * 5
                del tar[:]
                for i in range(outlength):
                    t = data[i + 5 + k]
                    tar.append(t)
                targ.extend(tar)
            p = 0
            for q in range(int(len(targ) / outlength)):
                target.insert(q, targ[p:p+5])
                p = p + 5
            return target
In [8]: # create shifted target sequence
        def shifted target(target, outlength):
            sh_target, target_in = list(), list()
            temp in = [0] * outlength
            for i in range(len(target)):
                templ = target[i]
                for j in range(1,5):
                    temp_in[j] = templ[j-1]
                target_in.extend(temp_in)
            for q in range(int(len(target_in) / outlength)):
                sh_target.insert(q, target_in[v:v+5])
                v = v + 5
```

```
In [9]: # create source items
         def create_source(data, outlength, target):
             source = list()
             isource = create_initial_source(data, outlength)
             source.append(isource)
             j = 1
             for i in range(int((len(data) / outlength)) - 2):
                 source.insert(j, target[i])
                 j = j + 1
             return source
In [10]: # create one-hot vectors for each number
         def create onehot(data, n features):
             hot list = to categorical(data, num classes=n features)
             return hot list
In [11]: # one-hot decoder
         def one hot decode(encode seq):
             return [argmax(vector) for vector in encode_seq]
In [12]: # prepare dataset for encoder-decoder models
         def create dataset(data, inlength, outlength, n features):
             # create target set
             target = create targets(data, outlength)
             # create shifted target set
             shift target = shifted target(target, outlength)
             # create source set
             source = create source(data, outlength, target)
             # one-hot encode data values
             target encoded = create onehot(target, n features)
             shift target encoded = create onehot(shift target, n features)
             source encoded = create onehot(source, n features)
             # define return values
             X1 = array(source encoded)
             X2 = array(shift target encoded)
             y = array(target encoded)
             return X1, X2, y
```

```
In [13]: # create your dataset for the encoder-decoder models
    inlength = 5  # input sequence length
    outlength = 5  # output sequence length
    cardinality = 39  # cardinality of the dataset
    n_features = cardinality + 1
    X1, X2, y = create_dataset(mylist, inlength, outlength, n_features)
    print("Source shape: ", X1.shape)
    print("Shifted target shape: ", X2.shape)
    print("Target shape: ", y.shape)

Source shape: (3632, 5, 40)
    Shifted target shape: (3632, 5, 40)
    Target shape: (3632, 5, 40)
```

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```
Source encoded:
0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0.
0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
Shifted target encoded:
[0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0.
Target encoded:
0. 0. 0. 0. 0. 1. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
_____
Source encoded: [7, 18, 19, 27, 30]
Shifted target encoded: [0, 12, 17, 25, 26]
Target encoded: [12, 17, 25, 26, 30]
```

```
In [15]: # create the training, inference encoder and inference decoder models
         def define models(n input, n output, n units):
             # define training encoder
             encoder inputs = Input(shape=(None, n input))
             encoder = LSTM(n_units, return_state=True)
             encoder_outputs, state_h, state_c = encoder(encoder_inputs)
             encoder_states = [state_h, state c]
             # define training decoder
             decoder inputs = Input(shape=(None, n output))
             decoder lstm = LSTM(n units, return sequences=True, return state=True)
             decoder_outputs, _, _ = decoder_lstm(decoder_inputs, initial_state=encoder_states)
             decoder dense = Dense(n output, activation='softmax')
             decoder outputs = decoder dense(decoder outputs)
             model = Model([encoder inputs, decoder inputs], decoder outputs)
             # define inference encoder
             encoder model = Model(encoder inputs, encoder states)
             # define inference decoder
             decoder state input h = Input(shape=(n units,))
             decoder_state_input_c = Input(shape=(n units,))
             decoder states inputs = [decoder state input h, decoder state input c]
             decoder_outputs, state_h, state_c = decoder_lstm(decoder_inputs, initial_state=decoder_states_inputs)
             decoder states = [state h, state c]
             decoder_outputs = decoder_dense(decoder outputs)
             decoder model = Model([decoder_inputs] + decoder_states_inputs, [decoder_outputs] + decoder_states)
             # return all models
             return model, encoder model, decoder model
```

```
In [16]: # generate target given source sequence
         def predict_sequence(infenc, infdec, source, n_steps, cardinality):
             # encode
             state = infenc.predict(source)
             # start of sequence input
             target_seq = array([0.0 for _ in range(cardinality)]).reshape(1,1, cardinality)
             # collect predictions
             output = list()
             for t in range(n_steps):
                 # predict the next number
                 yhat, h, c = infdec.predict([target seq] + state)
                 # store prediction
                 output.append(yhat[0,0,:])
                 # update state
                 state = [h, c]
                 # update target sequence
                 target_seq = yhat
             return array(output)
In [17]: # define the checkpoint
         filepath="weightsV2-{epoch:02d}-{acc:.4f}.hdf5"
         checkpoint = ModelCheckpoint(filepath, monitor='acc', verbose=1, save best only=True, mode='max')
         callbacks list = [checkpoint]
In [18]: # instantiate the model instance
         train, infenc, infdec = define models(n features, n features, 256)
         train.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['acc'])
```

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```
In [19]: nb_epochs = 500
train.fit([X1, X2], y, epochs=nb_epochs, callbacks=callbacks_list)
```

/Users/brent/anaconda3/lib/python3.6/site-packages/keras/engine/topology.py:2344: UserWarning: Layer lstm\_2 w as passed non-serializable keyword arguments: {'initial\_state': [<tf.Tensor 'lstm\_1/while/Exit\_2:0' shape=(?, 256) dtype=float32>, <tf.Tensor 'lstm\_1/while/Exit\_3:0' shape=(?, 256) dtype=float32>]}. They will not be inc luded in the serialized model (and thus will be missing at deserialization time).

str(node.arguments) + '. They will not be included '

```
Epoch 2/500
om 0.07087 to 0.09438, saving model to weights V2-02-0.0944.hdf5
Epoch 3/500
om 0.09438 to 0.10474, saving model to weights V2-03-0.1047.hdf5
Epoch 4/500
om 0.10474 to 0.11674, saving model to weightsV2-04-0.1167.hdf5
Epoch 5/500
om 0.11674 to 0.12572, saving model to weights V2-05-0.1257.hdf5
Epoch 6/500
om 0.12572 to 0.13320, saving model to weights V2-06-0.1332.hdf5
Epoch 7/500
om 0.13320 to 0.13827, saving model to weightsV2-07-0.1383.hdf5
Epoch 8/500
om 0.13827 to 0.13910, saving model to weightsV2-08-0.1391.hdf5
Epoch 9/500
om 0.13910 to 0.14295, saving model to weightsV2-09-0.1430.hdf5
Epoch 10/500
om 0.14295 to 0.15187, saving model to weights V2-10-0.1519.hdf5
Epoch 11/500
om 0.15187 to 0.15589, saving model to weights V2-11-0.1559.hdf5
Epoch 12/500
om 0.15589 to 0.15683, saving model to weights V2-12-0.1568.hdf5
Epoch 13/500
```

```
Out[19]: <keras.callbacks.History at 0x1a18b854a8>
In [20]: # load the best network weights
         filename = "weights-1137-0.9993.hdf5"
         train.load weights(filename)
         train.compile(optimizer='adam', loss='categorical crossentropy', metrics=['acc'])
In [201: # test model set
         src = [[2,5,6,12,27],[7,12,31,32,39],[4,23,25,30,39],[1,7,26,34,35],[3,4,24,31,35],[4,16,26,29,32],
         [1,17,23,26,35],[9,24,31,34,38],[12,18,25,32,34],[11,13,18,23,30]]
         correct = 0
         for m in range(len(src)-1):
             X3, , y1 = create dataset(src[m], inlength, outlength, n features)
             out sequence = predict sequence(infenc, infdec, X3, outlength, n features)
             target = one hot decode(out sequence)
             if (m < len(src)-1):
                 yhat = src[m+1]
             else: yhat = src[m]
             print('yhat: %s, prediction: %s' % (yhat, target))
             if np.array equal(yhat, target ):
                            correct += 1
         print("The prediction accuracy is: %.2f%%" % (float(correct)/float(len(src))*100.0))
         yhat: [7, 12, 31, 32, 39], prediction: [7, 12, 31, 32, 39]
         yhat: [4, 23, 25, 30, 39], prediction: [4, 23, 25, 30, 39]
         yhat: [1, 7, 26, 34, 35], prediction: [1, 7, 26, 34, 35]
         yhat: [3, 4, 24, 31, 35], prediction: [3, 4, 24, 31, 35]
         yhat: [4, 16, 26, 29, 32], prediction: [4, 16, 26, 29, 32]
         yhat: [1, 17, 23, 26, 35], prediction: [1, 17, 23, 26, 35]
         yhat: [9, 24, 31, 34, 38], prediction: [9, 24, 31, 34, 38]
         yhat: [12, 18, 25, 32, 34], prediction: [2, 10, 18, 23, 35]
         yhat: [11, 13, 18, 23, 30], prediction: [3, 4, 10, 13, 26]
         The prediction accuracy is: 70.00%
In [22]: # create new source input
         src = [9,24,31,34,38]
         X, , = create dataset(src, inlength, outlength, n features)
         out sequence = predict sequence(infenc, infdec, X, outlength, n features)
         output = one hot decode(out sequence)
         print("The predicted output sequence is: ", output)
         The predicted output sequence is: [4, 10, 21, 22, 35]
```

In [ ]: