

# 2\_pam\_model

May 12, 2019

## 0.1 Importing Required Packages

```
[0]: import numpy as np
import tensorflow as tf
from tensorflow import keras
import matplotlib.pyplot as plt
from tensorflow import keras
from tensorflow.keras.layers import *
from sklearn import preprocessing
import tensorflow.keras.backend as K
from sklearn.metrics import mean_squared_error
```

## 0.2 Hyper Parameters

```
[0]: msg_total = 2
channel = 4
epochs = 5000
sigma = 1e-4
batch_size = 1024
```

## 0.3 Defining required functions

```
[0]: def perturbation(x):
    w = K.random_normal(shape = (channel,2), mean=0.0,stddev=sigma**0.
    ↪5,dtype=None)
    xp = ((1-sigma)**0.5)*x + w
    return xp

def loss_tx(y_true, y_pred):
    return -y_true*y_pred

def get_policy(inp):
    xp = inp[0]
    x = inp[1]
    w = xp - x
    policy = -K.sum(w*w)
```

```
return policy
```

# 1 Transmitter

## 1.1 Defining Architecture

```
[0]: tx_inp = Input((1,))
      embeddings_layer = Dense(msg_total, activation = 'relu')(tx_inp)
      layer_dense = Dense(2*channel, activation = 'relu')(embeddings_layer)
      to_complex = Reshape((channel,2))(layer_dense)
      x = Lambda(lambda x: keras.backend.l2_normalize(x))(to_complex)
      xp = Lambda(perturbation)(to_complex)
      policy = Lambda(get_policy)([xp,x])
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/ops/resource\_variable\_ops.py:435: colocate\_with (from tensorflow.python.framework.ops) is deprecated and will be removed in a future version.

Instructions for updating:

Colocations handled automatically by placer.

## 1.2 Declaring Models

```
[0]: model_policy = keras.models.Model(inputs=tx_inp, outputs=policy)
      model_tx = keras.models.Model(inputs=tx_inp, outputs=xp)
      model_x = keras.models.Model(inputs=tx_inp, outputs=x)

      model_policy.compile(loss=loss_tx, optimizer=tf.keras.optimizers.SGD(lr = 1e-5))
      print(model_policy.summary())
```

```
-----
-----
Layer (type)                Output Shape           Param #   Connected to
=====
-----
input_1 (InputLayer)        (None, 1)              0         (None, 1)
-----
dense (Dense)                (None, 2)              4         input_1[0][0]
-----
dense_1 (Dense)              (None, 8)              24        dense[0][0]
-----
reshape (Reshape)           (None, 4, 2)           0         dense_1[0][0]
-----
```

```

-----
lambda_1 (Lambda)                (None, 4, 2)                0                reshape[0][0]
-----
-----
lambda (Lambda)                  (None, 4, 2)                0                reshape[0][0]
-----
-----
lambda_2 (Lambda)                ()                            0                lambda_1[0][0]
                                                                lambda[0][0]
=====
=====
Total params: 28
Trainable params: 28
Non-trainable params: 0
-----
-----
None

```

## 2 Receiver

### 2.1 Defining Architecture

```

[0]: rx_inp = Input((channel,2))
      to_flat = Reshape((2*channel,))(rx_inp)
      fc = Dense(5*2*channel, activation = 'relu')(to_flat)
      softmax = Dense(msg_total, activation = 'softmax')(fc)

      model_rx = keras.models.Model(inputs=rx_inp, outputs=softmax)

      model_rx.compile(loss=tf.keras.losses.categorical_crossentropy, optimizer=tf.
        ↳keras.optimizers.Adam())
      print(model_rx.summary())

```

```

-----
Layer (type)                 Output Shape                Param #
=====
input_2 (InputLayer)         (None, 4, 2)                0
-----
reshape_1 (Reshape)          (None, 8)                   0
-----
dense_2 (Dense)              (None, 40)                  360
-----
dense_3 (Dense)              (None, 2)                   82
=====
Total params: 442
Trainable params: 442
Non-trainable params: 0

```

-----  
None

## 2.2 Alternative Training

```
[0]: loss_tx = []
loss_rx = []
for epoch in range(epochs):
    # Transmitter training
    raw_input = np.random.randint(0,msg_total,(batch_size))
    label = np.zeros((batch_size, msg_total))
    label[np.arange(batch_size), raw_input] = 1
    tx_input = raw_input/float(msg_total)
    xp = model_tx.predict(tx_input)
    y = xp + np.random.normal(0,0.001,(batch_size, channel,2))
    pred = model_rx.predict(y)
    loss = np.sum(np.square(label - pred), axis = 1)
    history_tx = model_policy.fit(tx_input, loss, batch_size=batch_size,
    ↪epochs=1, verbose=0)
    loss_tx.append(history_tx.history['loss'][0])

    # Receiver Training
    raw_input = np.random.randint(0,msg_total,(batch_size))
    label = np.zeros((batch_size, msg_total))
    label[np.arange(batch_size), raw_input] = 1
    tx_input = raw_input/float(msg_total)
    x = model_x.predict(tx_input)
    y = x + np.random.normal(0,0.001,(batch_size, channel,2))
    history_rx = model_rx.fit(y, label, batch_size=batch_size, epochs=1,
    ↪verbose=0)
    loss_rx.append(history_rx.history['loss'][0])

    if(epoch % 100 == 0):
        print('epoch: ', epoch, 'tx_loss', history_tx.history['loss'][0],
        ↪'rx_loss', history_rx.history['loss'][0])
```

WARNING:tensorflow:From /usr/local/lib/python3.6/dist-packages/tensorflow/python/ops/math\_ops.py:3066: to\_int32 (from tensorflow.python.ops.math\_ops) is deprecated and will be removed in a future version.

Instructions for updating:

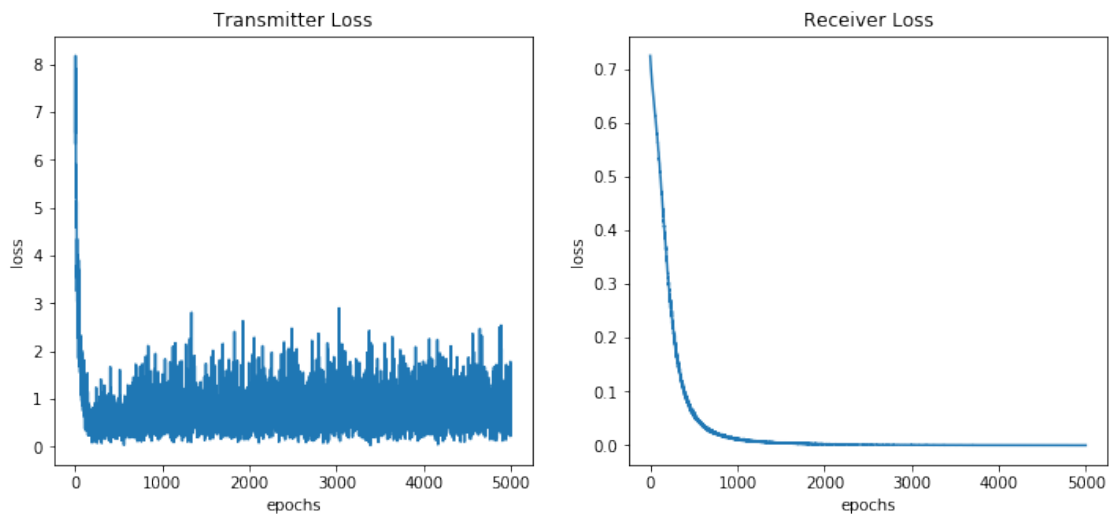
Use tf.cast instead.

```
epoch: 0 tx_loss 6.355036735534668 rx_loss 0.7242905497550964
epoch: 100 tx_loss 1.8137860298156738 rx_loss 0.5351258516311646
epoch: 200 tx_loss 0.8131426572799683 rx_loss 0.31978639960289
epoch: 300 tx_loss 0.4941194951534271 rx_loss 0.1713171899318695
epoch: 400 tx_loss 0.3641790747642517 rx_loss 0.09444549679756165
```

epoch: 500 tx\_loss 0.7153945565223694 rx\_loss 0.054864201694726944  
epoch: 600 tx\_loss 0.4647824764251709 rx\_loss 0.04075027257204056  
epoch: 700 tx\_loss 0.4751901626586914 rx\_loss 0.025828994810581207  
epoch: 800 tx\_loss 0.5345849990844727 rx\_loss 0.02043948322534561  
epoch: 900 tx\_loss 0.3673986792564392 rx\_loss 0.014428166672587395  
epoch: 1000 tx\_loss 0.3670782148838043 rx\_loss 0.011922061443328857  
epoch: 1100 tx\_loss 0.9511679410934448 rx\_loss 0.00840630941092968  
epoch: 1200 tx\_loss 0.6923137903213501 rx\_loss 0.007425171323120594  
epoch: 1300 tx\_loss 0.741775393486023 rx\_loss 0.00629937369376421  
epoch: 1400 tx\_loss 0.9444401860237122 rx\_loss 0.005360973533242941  
epoch: 1500 tx\_loss 1.7412992715835571 rx\_loss 0.004391422029584646  
epoch: 1600 tx\_loss 0.6625553369522095 rx\_loss 0.00392648670822382  
epoch: 1700 tx\_loss 0.8560501933097839 rx\_loss 0.0038145408034324646  
epoch: 1800 tx\_loss 0.7468227744102478 rx\_loss 0.0030808132141828537  
epoch: 1900 tx\_loss 0.7250442504882812 rx\_loss 0.002631203504279256  
epoch: 2000 tx\_loss 0.528317928314209 rx\_loss 0.002227142918854952  
epoch: 2100 tx\_loss 0.42298823595046997 rx\_loss 0.0019248031312599778  
epoch: 2200 tx\_loss 0.3724147081375122 rx\_loss 0.0018200164195150137  
epoch: 2300 tx\_loss 0.598807692527771 rx\_loss 0.0014720705803483725  
epoch: 2400 tx\_loss 0.570252537727356 rx\_loss 0.0015578294405713677  
epoch: 2500 tx\_loss 1.3871800899505615 rx\_loss 0.0013590501621365547  
epoch: 2600 tx\_loss 0.4627700746059418 rx\_loss 0.001310709398239851  
epoch: 2700 tx\_loss 1.3259549140930176 rx\_loss 0.001097007654607296  
epoch: 2800 tx\_loss 1.886702537536621 rx\_loss 0.0010604213457554579  
epoch: 2900 tx\_loss 1.357552409172058 rx\_loss 0.0008800771902315319  
epoch: 3000 tx\_loss 0.5879579782485962 rx\_loss 0.0009826362365856767  
epoch: 3100 tx\_loss 1.249929666519165 rx\_loss 0.0008248339290730655  
epoch: 3200 tx\_loss 0.49422529339790344 rx\_loss 0.0007180306711234152  
epoch: 3300 tx\_loss 0.7422264814376831 rx\_loss 0.0005964445881545544  
epoch: 3400 tx\_loss 1.3139021396636963 rx\_loss 0.0006083508487790823  
epoch: 3500 tx\_loss 0.5861632823944092 rx\_loss 0.0004838921595364809  
epoch: 3600 tx\_loss 0.1856534779071808 rx\_loss 0.0006139599718153477  
epoch: 3700 tx\_loss 0.494204044342041 rx\_loss 0.00048443354899063706  
epoch: 3800 tx\_loss 0.6790932416915894 rx\_loss 0.0004982298123650253  
epoch: 3900 tx\_loss 0.8209909200668335 rx\_loss 0.0003675811749417335  
epoch: 4000 tx\_loss 0.7746607065200806 rx\_loss 0.00046498404117301106  
epoch: 4100 tx\_loss 0.8726150989532471 rx\_loss 0.00035181891871616244  
epoch: 4200 tx\_loss 0.4192732572555542 rx\_loss 0.0003383115981705487  
epoch: 4300 tx\_loss 1.2453705072402954 rx\_loss 0.0003893995308317244  
epoch: 4400 tx\_loss 1.0274486541748047 rx\_loss 0.0003541375626809895  
epoch: 4500 tx\_loss 0.4736355245113373 rx\_loss 0.00022588712454307824  
epoch: 4600 tx\_loss 0.7548935413360596 rx\_loss 0.0003139497130177915  
epoch: 4700 tx\_loss 0.8765056133270264 rx\_loss 0.0002393309841863811  
epoch: 4800 tx\_loss 0.9064711332321167 rx\_loss 0.00023339706240221858  
epoch: 4900 tx\_loss 0.5094815492630005 rx\_loss 0.00019088402041234076

## 2.3 Plotting Transmitter and Receiver Losses

```
[0]: plt.figure(figsize = (12,5))
plt.subplot(1,2,1)
plt.plot(loss_tx)
plt.title('Transmitter Loss')
plt.xlabel('epochs')
plt.ylabel('loss')
plt.subplot(1,2,2)
plt.plot(loss_rx)
plt.title('Receiver Loss')
plt.xlabel('epochs')
plt.ylabel('loss')
plt.show()
```



## 2.4 Prediction

```
[0]: #testing
batch_size = 100
raw_input = np.random.randint(0,msg_total,(batch_size))
print(raw_input)
label = np.zeros((batch_size, msg_total))
label[np.arange(batch_size), raw_input] = 1
tx_input = raw_input/float(msg_total)
xp = model_x.predict(tx_input)
y = xp + np.random.normal(0,0.001,(batch_size, channel,2))
pred = model_rx.predict(y)
pred_int = np.argmax(pred, axis = 1)
print(pred_int)
```

```
from sklearn.metrics import accuracy_score  
  
print('accuracy:',accuracy_score(raw_input, pred_int))
```

```
[1 0 1 0 1 1 1 1 1 1 0 0 0 1 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 0 0 0 1 0 0  
 0 0 1 0 0 1 0 0 0 0 0 1 1 1 1 0 1 1 0 0 1 1 0 0 1 1 0 1 0 1 1 0 1 1 0 0 0  
 0 1 0 0 1 1 1 0 1 0 0 1 1 0 0 0 0 1 0 0 0 1 0 1 0 1]  
[1 0 1 0 1 1 1 1 1 1 0 0 0 1 1 0 0 1 0 1 1 0 0 0 1 0 0 1 0 0 0 0 0 0 1 0 0  
 0 0 1 0 0 1 0 0 0 0 0 1 1 1 1 0 1 1 0 0 1 1 0 0 1 1 0 1 0 1 1 0 1 1 0 0 0  
 0 1 0 0 1 1 1 0 1 0 0 1 1 0 0 0 0 1 0 0 0 1 0 1 0 1]  
accuracy: 1.0
```

[0]: