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## **CS505-01 Data Mining CRN-12050**

#### **Modify Parameters to The Models withe Neural Networks**

Pease submit the screen dumps

Showing the results

Make a Copy of the file:

tutorial05 different approaches to define neural networks keras.ipynb

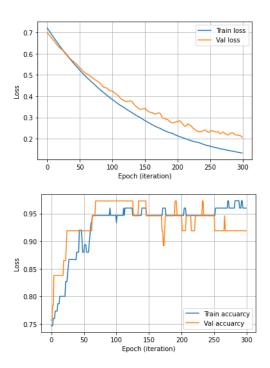
## 1.- Do the following modifications:

model.compile(optimizer='SGD',loss='categorical\_crossentropy', metrics=['accuracy']) # compiling the model

Train model and execute loss = model.evaluate(X\_test, Y\_oh\_test, verbose=0) print screen dumps

Training and Testing the Model model.compile(optimizer='SGD',loss='categorical\_crossentropy', metrics=['accuracy']) # compiling the model  $\label{eq:continuous} \mbox{\# training the model} \\ \mbox{history = model.fit}(X\_{train}, Y\_{oh\_{train}}, validation\_data=(X\_{val}, Y\_{oh\_{val}}), batch\_size= 64, epochs= 300)$ [113] V 25.5s Epoch 286/300 2/2 ------ 0s 36ms/step - accuracy: 0.8131 - loss: 0.4741 - val accuracy: 0.7297 - val loss: 0.5203 Enoch 287/300 --- 0s 71ms/step - accuracy: 0.8131 - loss: 0.4699 - val\_accuracy: 0.7027 - val\_loss: 0.5208 - 0s 36ms/step - accuracy: 0.8131 - loss: 0.4780 - val\_accuracy: 0.7297 - val\_loss: 0.5204 Epoch 289/300 — 0s 37ms/step - accuracy: 0.8078 - loss: 0.4761 - val\_accuracy: 0.7027 - val\_loss: 0.5203 ch 290/300 - 0s 36ms/step - accuracy: 0.8131 - loss: 0.4712 - val\_accuracy: 0.7297 - val\_loss: 0.5193 Epoch 291/300 Epoch 292/300 2/2 - 0s 37ms/step - accuracy: 0.8167 - loss: 0.4657 - val accuracy: 0.7297 - val loss: 0.5195 --- 0s 39ms/step - accuracy: 0.8235 - loss: 0.4606 - val accuracy: 0.7297 - val loss: 0.5174 Epoch 293/300 -- 0s 37ms/step - accuracy: 0.8219 - loss: 0.4720 - val\_accuracy: 0.7297 - val\_loss: 0.5174 Epoch 294/300 2/2 2/2 -9s 40ms/step - accuracy: 0.8272 - loss: 0.4600 - val\_accuracy: 0.7297 - val\_loss: 0.5175 Epoch 295/300 2/2 -—— 0s 38ms/step - accuracy: 0.8219 - loss: 0.4757 - val accuracy: 0.7297 - val loss: 0.5187 th 296/300 --- 0s 36ms/step - accuracy: 0.8183 - loss: 0.4658 - val\_accuracy: 0.7297 - val\_loss: 0.5181 2/2 -Epoch 298/300 --- 0s 37ms/step - accuracy: 0.8465 - loss: 0.4721 - val\_accuracy: 0.7297 - val\_loss: 0.5147 2/2 -- 0s 38ms/step - accuracy: 0.8308 - loss: 0.4633 - val\_accuracy: 0.7297 - val\_loss: 0.5152 Epoch 300/300 — 0s 35ms/step - accuracy: 0.8413 - loss: 0.4737 - val\_accuracy: 0.7297 - val\_loss: 0.5143

```
Test loss (cross-entropy and accuracy): [0.5236875414848328, 0.8157894611358643]
Layer 0
Bias:
[-0.0366995 -0.22425273 0.4229157 0.59154403 0.18444297]
[-0.19112627 0.75803596 -1.1145675 -1.0548539 -0.21111095]
 [ 0.5866124  0.7987467  -0.0984989  -0.2159796  0.7637586 ]]
Layer 1
 [ 0.08470635 -0.00670301 0.
                            -0.00695592 0.
                                              -0.29624605
 -0.01805937 0.
                  0.3159241 0.55739266]
-0.17694187 -0.17141363 0.33758232 -0.38800493]
 [-0.16499256 0.05249661 -0.29413685 0.3819801 0.16055244 0.37597921
 -0.5708797 -0.6043956 -0.78740287 -0.8950947 ]
  [ \ 0.09346582 \ -0.5408318 \ \ -0.49493673 \ -0.41793367 \ \ -0.51566553 \ \ -0.44807398 
  -0.15124473 -0.10380131 1.0826961 0.31763092]
 0.3023116 -0.48162222 0.93583125 0.6001836 ]
 [-0.372479 -0.07889292 -0.39420536 -0.62149185 -0.5120433 -0.19165218
  0.3711682 -0.2550603 -0.22121154 0.37638614]]
Layer 2
Bias:
[-1.1654588 0.23392682 0.9315328]
```



model.compile(optimizer='<u>RMSprop'</u>,loss='categorical\_crossentropy', metrics=['accuracy']) # compiling the model

 $\label{eq:continuous} Train\ model\ and\ execute\ loss = model.evaluate(X\_test,\ Y\_oh\_test,\ verbose=0)\ \ print\ screen\ dumps$ 

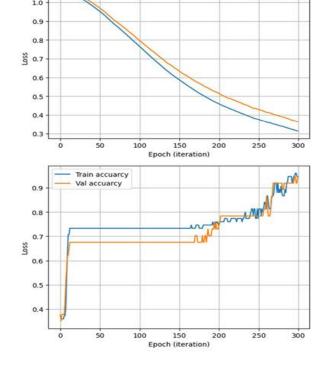
## Training and Testing the Model

[-1.7486603 0.16195932 1.4106075 ]

```
model.compile(optimizer='RMSprop',loss='categorical_crossentropy', metrics=['accuracy']) # compiling the model
                                                                                                                                                                            ▷₁ ▷↓ □ …
\triangleright
       # training the model
       history = model.fit(X_train, Y_oh_train, validation_data=(X_val,Y_oh_val),batch_size= 64, epochs= 300)

√ 26.4s

                                                                                                                                                                                         Python
  Epoch 286/300
    2/2 -
                          - 0s 37ms/step - accuracy: 1.0000 - loss: 0.2416 - val accuracy: 0.8919 - val loss: 0.2930
    Epoch 287/300
                         — Os 36ms/step - accuracy: 1.0000 - loss: 0.2416 - val_accuracy: 0.8919 - val_loss: 0.2950
    Epoch 288/300
    2/2 -
                         — 0s 35ms/step - accuracy: 1.0000 - loss: 0.2439 - val_accuracy: 0.8919 - val_loss: 0.2978
    Epoch 289/300
                         - 0s 34ms/step - accuracy: 1.0000 - loss: 0.2377 - val_accuracy: 0.8919 - val_loss: 0.2964
    2/2 -
    Epoch 290/300
    2/2 -
                         - 0s 48ms/step - accuracy: 1.0000 - loss: 0.2414 - val accuracy: 0.8919 - val loss: 0.2880
    Epoch 291/300
    2/2 -
                         — 0s 32ms/step - accuracy: 1.0000 - loss: 0.2355 - val_accuracy: 0.8919 - val_loss: 0.2906
    Epoch 292/300
    2/2 -
                          - 0s 33ms/step - accuracy: 1.0000 - loss: 0.2396 - val_accuracy: 0.9189 - val_loss: 0.2828
    Epoch 293/300
    2/2 -
                         - 0s 35ms/step - accuracy: 1.0000 - loss: 0.2392 - val_accuracy: 0.9189 - val_loss: 0.2822
    Epoch 294/300
    2/2 -
                          - 0s 34ms/step - accuracy: 1.0000 - loss: 0.2338 - val_accuracy: 0.8919 - val_loss: 0.2859
    Epoch 295/300
    2/2 -
                          0s 36ms/step - accuracy: 1.0000 - loss: 0.2341 - val_accuracy: 0.8919 - val_loss: 0.2883
    Epoch 296/300
                         — 0s 36ms/step - accuracy: 1.0000 - loss: 0.2347 - val_accuracy: 0.9189 - val_loss: 0.2815
    2/2 -
    Epoch 297/300
    2/2 -
                         - 0s 35ms/step - accuracy: 1.0000 - loss: 0.2286 - val_accuracy: 0.9189 - val_loss: 0.2749
    Epoch 298/300
    2/2 -
                          0s 37ms/step - accuracy: 1.0000 - loss: 0.2331 - val accuracy: 0.9189 - val loss: 0.2726
    Epoch 299/300
    2/2 -
                          0s 38ms/step - accuracy: 1.0000 - loss: 0.2332 - val_accuracy: 0.8919 - val_loss: 0.2811
    Epoch 300/300
                         — 0s 37ms/step - accuracy: 1.0000 - loss: 0.2280 - val_accuracy: 0.8919 - val_loss: 0.2878
   Test loss (cross-entropy and accuracy): [0.25814089179039, 1.0]
    [-0.13498312 -0.35740823 0.6129222 0.81119657 0.36762995]
    [[-0.14200217  0.13711931  0.38353345  0.10588858  0.16255262]
    [-0.2939697  0.23388055  1.0977377  0.70702153  -0.01464017]
    [-0.06110045 0.8323914 -1.2042543 -1.1110417 -0.13144428]
    [ 0.77725685  0.94617397 -0.24966045 -0.35755625  0.74486965]]
   Layer 1
   Bias:
    0.03121604 -0.00670301 0.
                                    -0.00879402 0.
                                                             -0.31457743
    -0.10699973 0.
                         0.47994897 0.7501659 ]
    -0.17694187 -0.17141363 0.33705753 -0.74605685]
    -0.5780685 -0.6043956 -0.9208475 -1.0565699 ]
    [ 0.53140587 -0.5408318 -0.49493673 -0.41793367 -0.51566553 -0.44807398
      0.23979351 -0.10380131 1.3729163 0.6028423 ]
    0.21388255 -0.48162222 1.4059473 1.0973687 ]
    [-0.53980386 \ -0.07889292 \ -0.39420536 \ -0.6230955 \ -0.5120433 \ -0.21000785
     0.29474616 -0.2550603 -0.21372405 0.36679882]]
   Layer 2
```

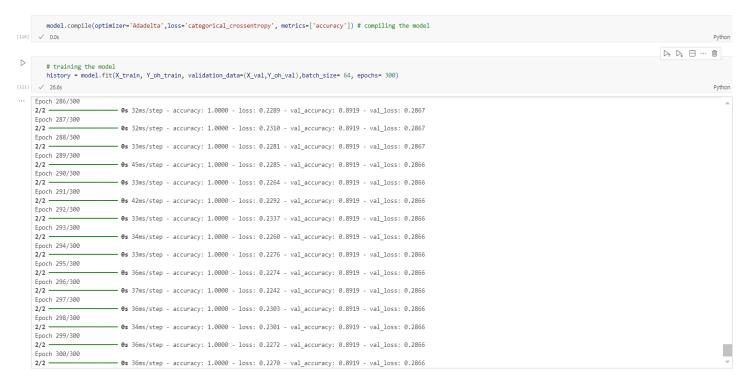


1.1

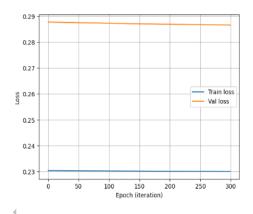
model.compile(optimizer='<u>Adadelta</u>',loss='categorical\_crossentropy', metrics=['accuracy']) # compiling the model

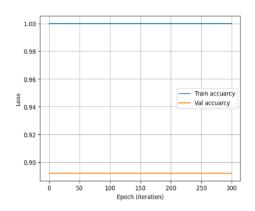
Train loss Val loss

Train model and execute loss = model.evaluate(X\_test, Y\_oh\_test, verbose=0) print screen dumps



```
Test loss (cross-entropy and accuracy): [0.2577400505542755, 1.0]
Layer 0
Bias:
[-0.13543561 -0.35772446 0.6132437 0.8115504 0.36791676]
[[-0.14224991 0.13696897 0.38362917 0.10602062 0.16289961]
[-0.2942179  0.2337426  1.0979195  0.7072452  -0.01466968]
[-0.06137059 0.83223677 -1.2041472 -1.1109128 -0.13108042]
[ 0.77709633  0.94608325 -0.24961124 -0.35749102  0.7451701 ]]
Layer 1
Bias:
[ 0.03126279 -0.00670301 0.
                              -0.00879402 0.
                                                     -0.31457743
-0.10699973 0.
                     0.4801719 0.7505655 ]
W:
[[\ 0.20699993\ -0.3681808\ \ 0.3854336\ \ -0.27194053\ \ 0.4504524\ \ -0.14875141
-0.17694187 -0.17141363 0.33697477 -0.7463477 ]
[-0.53096145 \quad 0.05249661 \quad -0.29413685 \quad 0.37977687 \quad 0.16055244 \quad 0.35762557
 -0.5780685 -0.6043956 -0.92086846 -1.0564437 ]
-0.23979351 -0.10380131 1.3733208 0.6033168 ]
0.21388255 -0.48162222 1.4066122 1.0980369 ]
[-0.53987586 \ -0.07889292 \ -0.39420536 \ -0.6230955 \ -0.5120433 \ -0.21000785
 0.29474616 -0.2550603 -0.2136348  0.36708045]]
Layer 2
Bias:
[-1.7496338 0.1616593 1.4113752]
```

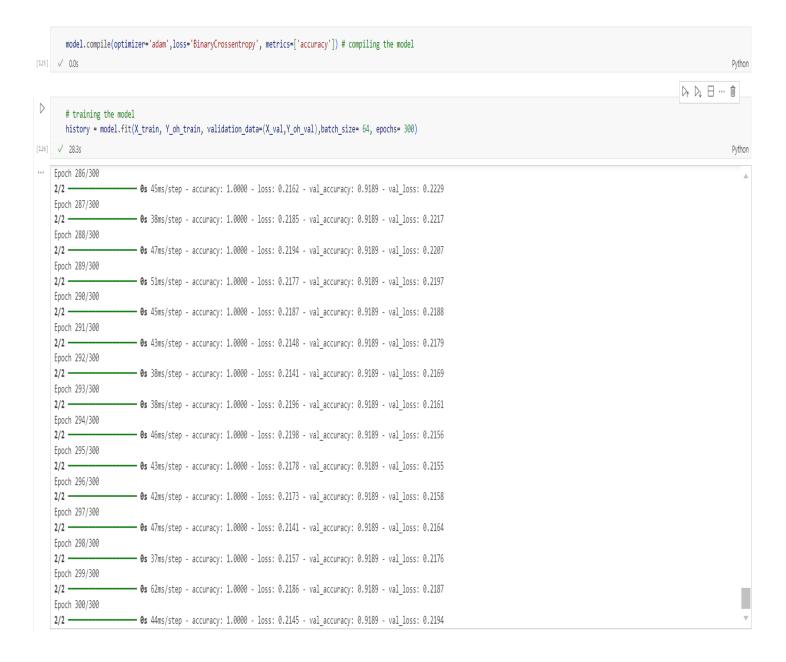




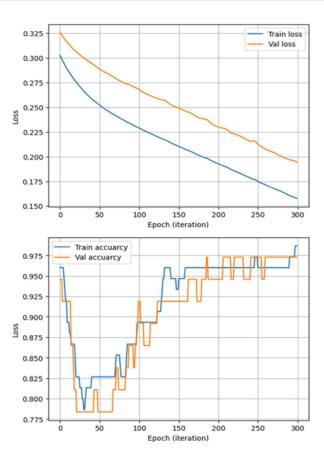
## 2. - Do the following Modifications

model.compile(optimizer='adam',loss='<u>BinaryCrossentropy'</u>, metrics=['accuracy']) # compiling the model

Train model and execute loss = model.evaluate(X\_test, Y\_oh\_test, verbose=0) print screen dumps

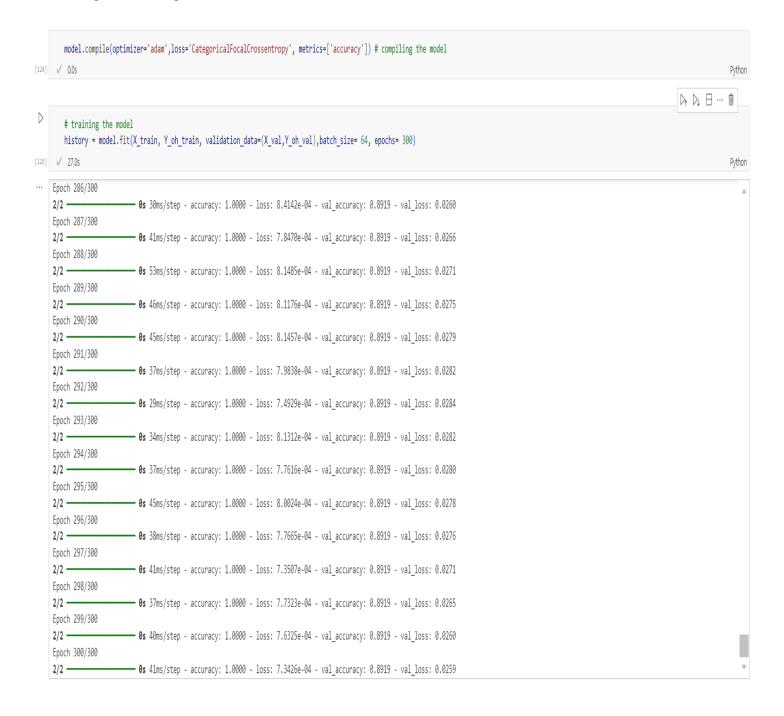


```
Test loss (cross-entropy and accuracy): [0.22904737293720245, 1.0]
Layer 0
Bias:
[-0.28949985 -0.45642433 0.7236772 1.0158486 0.2676207 ]
[[-0.07670111 0.15791214 0.3446164 0.06825446 0.29289162]
[ 0.00888288  0.8567143  -1.2425574  -1.1933881  0.1887195 ]
Layer 1
[ 0.17568761 -0.00670301 0.
                       -0.00879402 0.
                                      -0.31457743
-0.10699973 0. 0.43360683 0.7571448 ]
-0.17694187 -0.17141363 0.19597746 -1.2533767 ]
-0.5780685 -0.6043956 -0.8379493 -1.0462577 ]
-0.23979351 -0.10380131 1.2143216 0.46561015]
0.21388255 -0.48162222 1.2621231 0.99175054]
[-0.6976439 \quad -0.07889292 \quad -0.39420536 \quad -0.6230955 \quad -0.5120433 \quad -0.21000785
 0.29474616 -0.2550603 -0.19222246 0.40559667]]
Layer 2
Bias:
[-2.2561243 -0.23163246 1.5753326 ]
```

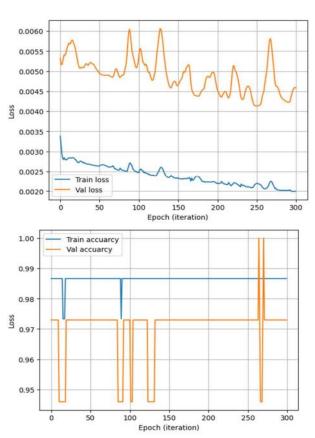


model.compile(optimizer='adam', loss='CategoricalFocalCrossentropy', metrics=['accuracy']) # compiling the model

Train model and execute loss = model.evaluate(X\_test, Y\_oh\_test, verbose=0) print screen dumps



```
Test loss (cross-entropy and accuracy): [0.002581462264060974, 1.0]
Layer 0
Bias:
[-0.29694796 -0.4679683  0.740621  1.0291415  0.28251413]
W:
[[-0.05672254 0.18002853 0.32165316 0.04486754 0.29037398]
[-0.28460047 0.23922607 1.1883373 0.95583576 -0.22361849]
[ 0.05654532  0.89411664 -1.2742827 -1.229683  0.15152498]
Layer 1
Bias:
[ 0.15409765 -0.00670301 0.
                          -0.00879402 0.
                                            -0.31457743
-0.10699973 0.
                0.46120867 0.7701815 ]
-0.17694187 -0.17141363 0.12114882 -1.3880554 ]
-0.5780685 -0.6043956 -0.8594622 -1.0913069 ]
[ 0.8510568 -0.5408318 -0.49493673 -0.41793367 -0.51566553 -0.44807398
 -0.23979351 -0.10380131 1.3465645 0.6298507 ]
0.21388255 -0.48162222 1.4252068 1.1754106 ]
[-0.7479442 -0.07889292 -0.39420536 -0.6230955 -0.5120433 -0.21000785
  0.29474616 -0.2550603 -0.19446278 0.3826001 ]]
Layer 2
Bias:
[-2.612462 -0.3135117 1.807699]
```



model.compile(optimizer='adam',loss='SparseCategoricalCrossentropy', metrics=['accuracy']) # compiling the model

Train model and execute loss = model.evaluate(X\_test, Y\_oh\_test, verbose=0) print screen dumps

```
model.compile(optimizer='adam',loss='SparseCategoricalCrossentropy', metrics=['accuracy']) # compiling the model
        # training the model
        history = model.fit(X_train, Y_oh_train, validation_data=(X_val,Y_oh_val),batch_size= 64, epochs= 300)
[132] × 0.4s
     Cell In[132], <u>line 2</u>
           1 # training the model
     ----> 2 history = model.fit(X_train, Y_oh_train, validation_data=(X_val,Y_oh_val),batch_size= 64, epochs= 300)
     File ~\AppData\Roaming\Python\Python311\site-packages\keras\src\utils\traceback_utils.py:122, in filter_traceback.<locals>.error_handler(*args, **kwargs)
         filtered_tb = _process_traceback_frames(e.__traceback_)
# To get the full stack trace, call:
# `keras.config.disable_traceback_filtering()`
                 raise e.with_traceback(filtered_tb) from None
         123 finally:
124 del filtered_tb
     File ~\AppData\Roaming\Python\Python311\site-packages\keras\src\backend\tensorflow\nn.py:652, in sparse_categorical_crossentropy(target, output, from_logits, axis)
                 raise ValueError(
         647
                      "Argument `output` must be at least rank 1. "
         649
                      f"output.shape={output.shape}"
         650
         651 if len(target.shape) != len(output.shape[:-1]):
                 raise ValueError(
                      "Argument `output` must have rank (ndim) `target.ndim - 1`. " \tt "Received: " \tt "
         653
         <u>655</u>
                      f"target.shape={target.shape}, output.shape={output.shape}"
         656
         657 for e1, e2 in zip(target.shape, output.shape[:-1]):
                 if e1 is not None and e2 is not None and e1 != e2:
     ValueError: Argument `output` must have rank (ndim) `target.ndim - 1`. Received: target.shape=(None, 3), output.shape=(None, 3)
```