

AutoML Models for Modulation Classification

March 24, 2021

1 Modulation Classification

Developing AutoML Model for AWGN Channel for Modulation Classification. SNR Ratio's(in dB) of Channel are [5, 10, 15, 20, 25, 30].

Data Generation: Data is generated using MatLab's Communication Toolbox for Modulation Schemes QPSK, 16-QAM, 64-QAM for SNR Ratio's (in dB) [5, 10, 15, 20, 25, 30] when signal is transmitted through Channels AWGN and Rayleigh Channels.

Modulation Classification: We will use AutoML to create a classifier that predicts Modulation Scheme depending on In-Phase and Quadrature-Phase Components at the Receiver's End.

1.1 Imports

1.1.1 Importing Libraries

We will be using AutoKeras for generating AutoML Models. Source of Documentation: <https://autokeras.com/>

```
[1]: import numpy as np
import matplotlib.pyplot as plt
import matplotlib.image as pimg
import seaborn as sns
import scipy.io
import os

# AutoML Libraries
!pip3 install autokeras
import autokeras as ak

# Tensorflow Libraries
import tensorflow as tf
```

Requirement already satisfied: autokeras in /usr/local/lib/python3.7/dist-packages (1.0.12)

Requirement already satisfied: pandas in /usr/local/lib/python3.7/dist-packages (from autokeras) (1.1.5)

Requirement already satisfied: keras-tuner>=1.0.2 in

/usr/local/lib/python3.7/dist-packages (from autokeras) (1.0.2)
Requirement already satisfied: packaging in /usr/local/lib/python3.7/dist-packages (from autokeras) (20.9)
Requirement already satisfied: scikit-learn in /usr/local/lib/python3.7/dist-packages (from autokeras) (0.22.2.post1)
Requirement already satisfied: tensorflow>=2.3.0 in /usr/local/lib/python3.7/dist-packages (from autokeras) (2.4.1)
Requirement already satisfied: numpy>=1.15.4 in /usr/local/lib/python3.7/dist-packages (from pandas->autokeras) (1.19.5)
Requirement already satisfied: python-dateutil>=2.7.3 in /usr/local/lib/python3.7/dist-packages (from pandas->autokeras) (2.8.1)
Requirement already satisfied: pytz>=2017.2 in /usr/local/lib/python3.7/dist-packages (from pandas->autokeras) (2018.9)
Requirement already satisfied: colorama in /usr/local/lib/python3.7/dist-packages (from keras-tuner>=1.0.2->autokeras) (0.4.4)
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Requirement already satisfied: joblib>=0.11 in /usr/local/lib/python3.7/dist-packages (from scikit-learn->autokeras) (1.0.1)
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Requirement already satisfied: h5py~=2.10.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow>=2.3.0->autokeras) (2.10.0)
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Requirement already satisfied: wrapt~=1.12.1 in /usr/local/lib/python3.7/dist-
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Requirement already satisfied: absl-py~=0.10 in /usr/local/lib/python3.7/dist-
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Requirement already satisfied: protobuf>=3.9.2 in /usr/local/lib/python3.7/dist-
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Requirement already satisfied: gast==0.3.3 in /usr/local/lib/python3.7/dist-
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Requirement already satisfied: tensorflow-estimator<2.5.0,>=2.4.0 in
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(2.4.0)
Requirement already satisfied: idna<3,>=2.5 in /usr/local/lib/python3.7/dist-
packages (from requests->keras-tuner>=1.0.2->autokeras) (2.10)
Requirement already satisfied: urllib3!=1.25.0,!1.25.1,<1.26,>=1.21.1 in
/usr/local/lib/python3.7/dist-packages (from requests->keras-
tuner>=1.0.2->autokeras) (1.24.3)
Requirement already satisfied: certifi>=2017.4.17 in
/usr/local/lib/python3.7/dist-packages (from requests->keras-
tuner>=1.0.2->autokeras) (2020.12.5)
Requirement already satisfied: chardet<4,>=3.0.2 in
/usr/local/lib/python3.7/dist-packages (from requests->keras-
tuner>=1.0.2->autokeras) (3.0.4)
Requirement already satisfied: setuptools>=41.0.0 in
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tensorboard~=2.4->tensorflow>=2.3.0->autokeras) (54.1.2)
Requirement already satisfied: werkzeug>=0.11.15 in
/usr/local/lib/python3.7/dist-packages (from
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Requirement already satisfied: google-auth-oauthlib<0.5,>=0.4.1 in
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 Requirement already satisfied: markdown>=2.6.8 in /usr/local/lib/python3.7/dist-
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 Requirement already satisfied: rsa<5,>=3.1.4; python_version >= "3.6" in
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 auth<2,>=1.6.3->tensorboard~=2.4->tensorflow>=2.3.0->autokeras) (4.7.2)
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 oauthlib<0.5,>=0.4.1->tensorboard~=2.4->tensorflow>=2.3.0->autokeras) (1.3.0)
 Requirement already satisfied: importlib-metadata; python_version < "3.8" in
 /usr/local/lib/python3.7/dist-packages (from
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 Requirement already satisfied: pyasn1<0.5.0,>=0.4.6 in
 /usr/local/lib/python3.7/dist-packages (from pyasn1-modules>=0.2.1->google-
 auth<2,>=1.6.3->tensorboard~=2.4->tensorflow>=2.3.0->autokeras) (0.4.8)
 Requirement already satisfied: oauthlib>=3.0.0 in /usr/local/lib/python3.7/dist-
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 Requirement already satisfied: zipp>=0.5 in /usr/local/lib/python3.7/dist-
 packages (from importlib-metadata; python_version <
 "3.8"->markdown>=2.6.8->tensorboard~=2.4->tensorflow>=2.3.0->autokeras) (3.4.1)

1.1.2 Importing Data

Files are uploaded in Google Drive. Notebook is connected to Google Drive

```
[2]: from google.colab import drive
drive.mount('/gdrive')
%cd /gdrive/My Drive/Modulation-Classification/AutoML
```

Drive already mounted at /gdrive; to attempt to forcibly remount, call
 drive.mount("/gdrive", force_remount=True).
 /gdrive/My Drive/Modulation-Classification/AutoML

1.2 Processing Data

Importing Data for all SNR Ratio's

```
[3]: def ImportData(Path):
    ModulationSchemes = os.listdir(Path)

    Dataset = {}
    for scheme in ModulationSchemes:
        DataPath = Path + "/" + scheme
        files = os.listdir(DataPath)
        DataofScheme = {}
        for f in files:
            Data = scipy.io.loadmat(DataPath + "/" + f)
            snr = Data['snr'][0][0]
            rx = Data['rx']
            InPhase, QuadPhase = rx.real, rx.imag
            DataofScheme[snr] = np.append(InPhase, QuadPhase, axis=1)
        Dataset[scheme] = DataofScheme

    return Dataset
```

1.3 Train and Validation Datasets

Training: - Received Signal with SNR Ratio 30dB is used for Training for both Channels and all Modulation Schemes.

Validation: - AutoML is validated on Received Signals with SNR Ratio's (in dB) [5, 10, 15, 20, 25]

```
[4]: def GenerateDatasets(Channel, L=None):
    if Channel == "AWGN":
        Path = "../Data/" + Channel
    elif Channel == "Rayleigh":
        Path = "../Data/" + Channel + "/" + str(L)
    Data = ImportData(Path)

    Dataset = {}
    Dataset['Classes'] = list(Data.keys())
    OneHotClasses = np.eye(len(Dataset['Classes']))

    Classes = {}
    for i in range(len(Dataset['Classes'])):
        Classes[Dataset['Classes'][i]] = OneHotClasses[i]

    Valid_SNRs = [5, 10, 15, 20, 25]

    if Channel == "AWGN":
        X_Train, y_Train = np.empty((0, 2)), np.empty((0, 3))
        X_Valid, y_Valid = {}, {}

        for snr in Valid_SNRs:
```

```

X_Valid[snr] = np.empty((0,2))
y_Valid[snr] = np.empty((0,3))

for c in Classes.keys():
    ModData = Data[c]
    SNRs = ModData.keys()
    for snr in SNRs:
        if snr == 30:
            X = ModData[snr]
            y = np.repeat(np.expand_dims(Classes[c],axis=0), X.
↪shape[0], axis=0)
            X_Train = np.append(X_Train,X,axis=0)
            y_Train = np.append(y_Train,y,axis=0)
        else:
            X = ModData[snr]
            y = np.repeat(np.expand_dims(Classes[c],axis=0), X.
↪shape[0], axis=0)
            X_Valid[snr] = np.append(X_Valid[snr], X, axis=0)
            y_Valid[snr] = np.append(y_Valid[snr], y, axis=0)

elif Channel == "Rayleigh":
    X_Train, y_Train = np.empty((0,100,2)), np.empty((0,3))
    X_Valid, y_Valid = {}, {}
    for snr in Valid_SNRs:
        X_Valid[snr] = np.empty((0,100,2))
        y_Valid[snr] = np.empty((0,3))

    for c in Classes.keys():
        ModData = Data[c]
        SNRs = ModData.keys()
        for snr in SNRs:
            if snr == 30:
                X = ModData[snr]
                X = X.reshape(-1,100,2)
                y = np.repeat(np.expand_dims(Classes[c],axis=0), X.
↪shape[0], axis=0)
                X_Train = np.append(X_Train,X,axis=0)
                y_Train = np.append(y_Train,y,axis=0)
            else:
                X = ModData[snr]
                X = X.reshape(-1,100,2)
                y = np.repeat(np.expand_dims(Classes[c],axis=0), X.
↪shape[0], axis=0)
                X_Valid[snr] = np.append(X_Valid[snr], X, axis=0)
                y_Valid[snr] = np.append(y_Valid[snr], y, axis=0)

```

```
return X_Train, y_Train, X_Valid, y_Valid
```

1.4 Evaluating Data

```
[5]: def EvaluateData(Model, X_Train, y_Train, X_Valid, y_Valid, Name):
    Valid_SNR = np.array([5,10,15,20,25,30])
    Accuracy = []

    print ("Evaluating Model")
    for snr in Valid_SNR:
        if snr == 30:
            Loss, Acc = Model.evaluate(X_Train, y_Train)
        else:
            Loss, Acc = Model.evaluate(X_Valid[snr], y_Valid[snr])
        print ("SNR:", snr, "Accuracy:", Acc)
        Accuracy.append(Acc)

    Accuracy = np.array(Accuracy)

    plt.figure(figsize=(10,10))
    plt.plot(Valid_SNR,Accuracy, color='blue')
    plt.scatter(Valid_SNR,Accuracy, color='red')
    plt.title("Accuracy vs SNR")
    plt.xlabel("SNR")
    plt.ylabel("Accuracy")
    plt.grid()
    plt.savefig("Images/" + Name)
    plt.show()
```

1.5 AutoML Model for AWGN Channel

For AWGN Channel, Input Dimensions is (2,)

AWGN Data

```
[6]: X_Train, y_Train, X_Valid, y_Valid = GenerateDatasets('AWGN')
```

Creating a Classifier

```
[7]: AWGNClassifier = ak.StructuredDataClassifier(
    overwrite=True,
    max_trials=10,
    loss='categorical_crossentropy',
    num_classes=3)
```

```
AWGNClassifier.fit(X_Train, y_Train, epochs=15, batch_size=64,  
↳validation_split=0.1)
```

Trial 10 Complete [00h 00m 19s]

val_accuracy: 1.0

Best val_accuracy So Far: 1.0

Total elapsed time: 00h 03m 05s

INFO:tensorflow:Oracle triggered exit

Epoch 1/15

469/469 [=====] - 2s 2ms/step - loss: 0.9269 -

accuracy: 0.8846

Epoch 2/15

469/469 [=====] - 1s 2ms/step - loss: 1.2889 -

accuracy: 0.4880

Epoch 3/15

469/469 [=====] - 1s 2ms/step - loss: 0.8460 -

accuracy: 0.6703

Epoch 4/15

469/469 [=====] - 1s 2ms/step - loss: 0.7505 -

accuracy: 0.7085

Epoch 5/15

469/469 [=====] - 1s 2ms/step - loss: 0.6694 -

accuracy: 0.7330

Epoch 6/15

469/469 [=====] - 1s 2ms/step - loss: 0.6282 -

accuracy: 0.7421

Epoch 7/15

469/469 [=====] - 1s 2ms/step - loss: 0.5567 -

accuracy: 0.7586

Epoch 8/15

469/469 [=====] - 1s 2ms/step - loss: 0.5165 -

accuracy: 0.7678

Epoch 9/15

469/469 [=====] - 1s 2ms/step - loss: 0.5981 -

accuracy: 0.7578

Epoch 10/15

469/469 [=====] - 1s 2ms/step - loss: 0.4665 -

accuracy: 0.7864

Epoch 11/15

469/469 [=====] - 1s 2ms/step - loss: 0.4329 -

accuracy: 0.7991

Epoch 12/15

469/469 [=====] - 1s 2ms/step - loss: 0.3973 -

accuracy: 0.8129

Epoch 13/15

469/469 [=====] - 1s 2ms/step - loss: 0.3581 -


```

accuracy: 0.8296
Epoch 14/15
469/469 [=====] - 1s 2ms/step - loss: 0.3219 -
accuracy: 0.8450
Epoch 15/15
469/469 [=====] - 1s 2ms/step - loss: 0.2883 -
accuracy: 0.8602
INFO:tensorflow:Assets written to:
./structured_data_classifier/best_model/assets

```

Model Summary

```

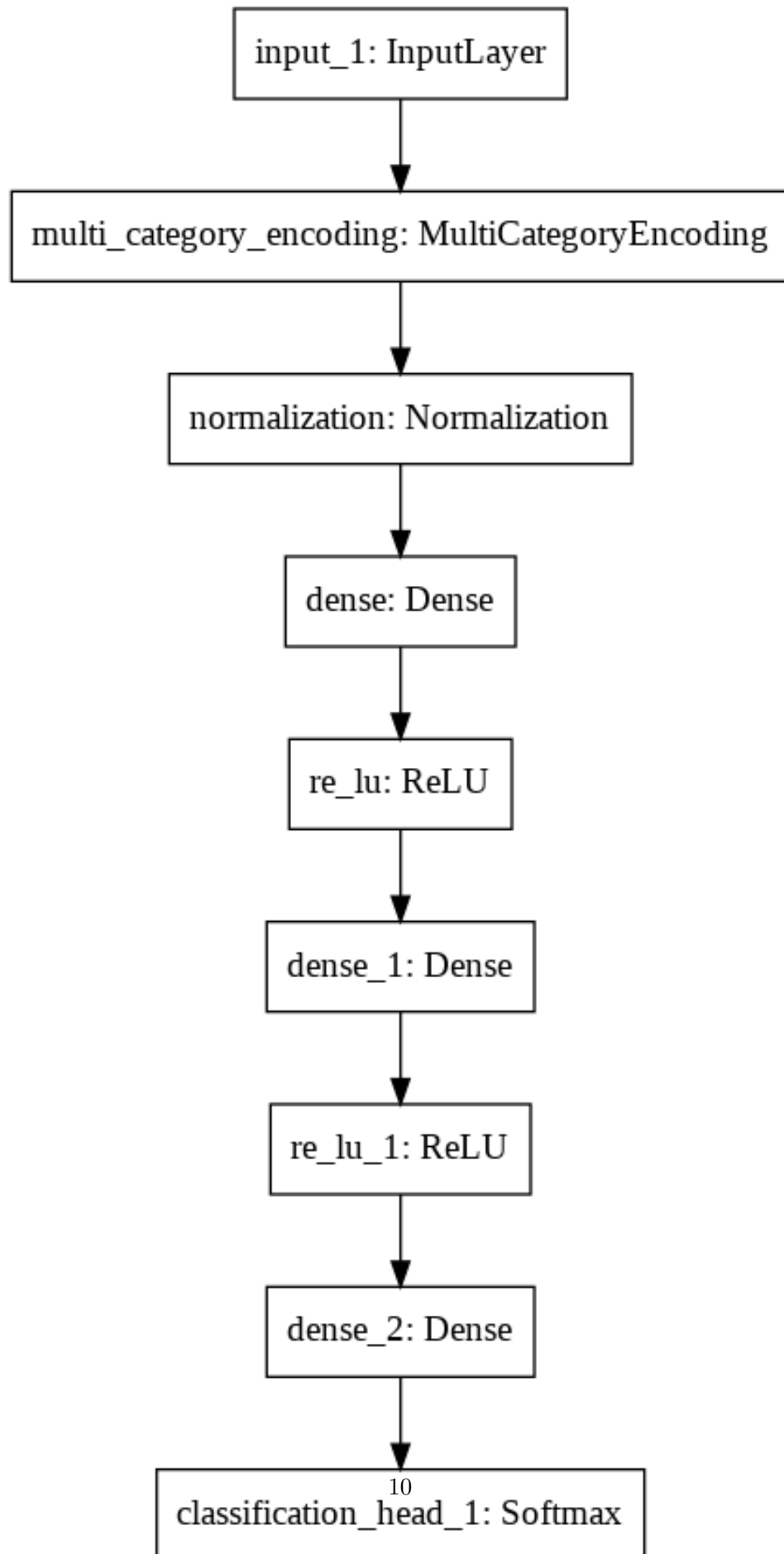
[8]: AWGN_Model = AWGNClassifier.export_model()
      AWGN_Model.summary()
      tf.keras.utils.plot_model(AWGN_Model, to_file='Images/AWGN_Model.png',
      ↪show_shapes=False,show_layer_names=True)

```

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 2)]	0
multi_category_encoding (Mul	(None, 2)	0
normalization (Normalization	(None, 2)	5
dense (Dense)	(None, 32)	96
re_lu (ReLU)	(None, 32)	0
dense_1 (Dense)	(None, 32)	1056
re_lu_1 (ReLU)	(None, 32)	0
dense_2 (Dense)	(None, 3)	99
classification_head_1 (Softm	(None, 3)	0
Total params: 1,256		
Trainable params: 1,251		
Non-trainable params: 5		

[8]:



1.5.1 Training and Evaluating Model

Training the Model

```
[9]: AWGN_Model.fit(X_Train, y_Train, epochs=50, batch_size=32, validation_split=0.1, shuffle=True)
```

Epoch 1/50

844/844 [=====] - 2s 3ms/step - loss: 0.2908 - accuracy: 0.8854 - val_loss: 0.3373 - val_accuracy: 1.0000

Epoch 2/50

844/844 [=====] - 2s 2ms/step - loss: 0.2221 - accuracy: 0.9057 - val_loss: 0.3141 - val_accuracy: 1.0000

Epoch 3/50

844/844 [=====] - 2s 2ms/step - loss: 0.2115 - accuracy: 0.9062 - val_loss: 0.3142 - val_accuracy: 1.0000

Epoch 4/50

844/844 [=====] - 2s 2ms/step - loss: 0.2075 - accuracy: 0.9066 - val_loss: 0.2998 - val_accuracy: 1.0000

Epoch 5/50

844/844 [=====] - 2s 2ms/step - loss: 0.2063 - accuracy: 0.9065 - val_loss: 0.3199 - val_accuracy: 1.0000

Epoch 6/50

844/844 [=====] - 2s 2ms/step - loss: 0.2055 - accuracy: 0.9065 - val_loss: 0.2924 - val_accuracy: 1.0000

Epoch 7/50

844/844 [=====] - 2s 2ms/step - loss: 0.2054 - accuracy: 0.9064 - val_loss: 0.2741 - val_accuracy: 1.0000

Epoch 8/50

844/844 [=====] - 2s 2ms/step - loss: 0.2052 - accuracy: 0.9065 - val_loss: 0.3102 - val_accuracy: 1.0000

Epoch 9/50

844/844 [=====] - 2s 2ms/step - loss: 0.2049 - accuracy: 0.9065 - val_loss: 0.3277 - val_accuracy: 1.0000

Epoch 10/50

844/844 [=====] - 2s 2ms/step - loss: 0.2047 - accuracy: 0.9064 - val_loss: 0.3357 - val_accuracy: 1.0000

Epoch 11/50

844/844 [=====] - 2s 2ms/step - loss: 0.2048 - accuracy: 0.9065 - val_loss: 0.3179 - val_accuracy: 1.0000

Epoch 12/50

844/844 [=====] - 2s 2ms/step - loss: 0.2044 - accuracy: 0.9066 - val_loss: 0.3423 - val_accuracy: 1.0000

Epoch 13/50

844/844 [=====] - 2s 2ms/step - loss: 0.2044 -

```

accuracy: 0.9065 - val_loss: 0.3226 - val_accuracy: 1.0000
Epoch 14/50
844/844 [=====] - 2s 2ms/step - loss: 0.2046 -
accuracy: 0.9066 - val_loss: 0.3269 - val_accuracy: 1.0000
Epoch 15/50
844/844 [=====] - 2s 2ms/step - loss: 0.2046 -
accuracy: 0.9065 - val_loss: 0.2747 - val_accuracy: 1.0000
Epoch 16/50
844/844 [=====] - 2s 2ms/step - loss: 0.2043 -
accuracy: 0.9066 - val_loss: 0.3025 - val_accuracy: 1.0000
Epoch 17/50
844/844 [=====] - 2s 2ms/step - loss: 0.2044 -
accuracy: 0.9066 - val_loss: 0.2811 - val_accuracy: 1.0000
Epoch 18/50
844/844 [=====] - 2s 2ms/step - loss: 0.2044 -
accuracy: 0.9064 - val_loss: 0.3501 - val_accuracy: 1.0000
Epoch 19/50
844/844 [=====] - 2s 2ms/step - loss: 0.2045 -
accuracy: 0.9066 - val_loss: 0.3346 - val_accuracy: 1.0000
Epoch 20/50
844/844 [=====] - 2s 2ms/step - loss: 0.2043 -
accuracy: 0.9065 - val_loss: 0.3080 - val_accuracy: 1.0000
Epoch 21/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9066 - val_loss: 0.3266 - val_accuracy: 1.0000
Epoch 22/50
844/844 [=====] - 2s 2ms/step - loss: 0.2043 -
accuracy: 0.9061 - val_loss: 0.2784 - val_accuracy: 1.0000
Epoch 23/50
844/844 [=====] - 2s 2ms/step - loss: 0.2041 -
accuracy: 0.9066 - val_loss: 0.2737 - val_accuracy: 1.0000
Epoch 24/50
844/844 [=====] - 2s 2ms/step - loss: 0.2045 -
accuracy: 0.9064 - val_loss: 0.3054 - val_accuracy: 1.0000
Epoch 25/50
844/844 [=====] - 2s 2ms/step - loss: 0.2042 -
accuracy: 0.9066 - val_loss: 0.3233 - val_accuracy: 1.0000
Epoch 26/50
844/844 [=====] - 2s 2ms/step - loss: 0.2041 -
accuracy: 0.9065 - val_loss: 0.3178 - val_accuracy: 1.0000
Epoch 27/50
844/844 [=====] - 2s 2ms/step - loss: 0.2041 -
accuracy: 0.9066 - val_loss: 0.2944 - val_accuracy: 1.0000
Epoch 28/50
844/844 [=====] - 2s 2ms/step - loss: 0.2042 -
accuracy: 0.9065 - val_loss: 0.3252 - val_accuracy: 1.0000
Epoch 29/50
844/844 [=====] - 2s 2ms/step - loss: 0.2044 -

```

```

accuracy: 0.9066 - val_loss: 0.3275 - val_accuracy: 1.0000
Epoch 30/50
844/844 [=====] - 2s 2ms/step - loss: 0.2042 -
accuracy: 0.9064 - val_loss: 0.3146 - val_accuracy: 1.0000
Epoch 31/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9065 - val_loss: 0.2944 - val_accuracy: 0.9997
Epoch 32/50
844/844 [=====] - 2s 2ms/step - loss: 0.2043 -
accuracy: 0.9064 - val_loss: 0.3273 - val_accuracy: 0.9937
Epoch 33/50
844/844 [=====] - 2s 2ms/step - loss: 0.2044 -
accuracy: 0.9064 - val_loss: 0.2853 - val_accuracy: 1.0000
Epoch 34/50
844/844 [=====] - 2s 2ms/step - loss: 0.2041 -
accuracy: 0.9064 - val_loss: 0.2936 - val_accuracy: 1.0000
Epoch 35/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9067 - val_loss: 0.3164 - val_accuracy: 1.0000
Epoch 36/50
844/844 [=====] - 2s 2ms/step - loss: 0.2041 -
accuracy: 0.9066 - val_loss: 0.3026 - val_accuracy: 1.0000
Epoch 37/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9064 - val_loss: 0.2883 - val_accuracy: 1.0000
Epoch 38/50
844/844 [=====] - 2s 2ms/step - loss: 0.2042 -
accuracy: 0.9064 - val_loss: 0.2828 - val_accuracy: 1.0000
Epoch 39/50
844/844 [=====] - 2s 2ms/step - loss: 0.2042 -
accuracy: 0.9065 - val_loss: 0.3158 - val_accuracy: 1.0000
Epoch 40/50
844/844 [=====] - 2s 3ms/step - loss: 0.2041 -
accuracy: 0.9065 - val_loss: 0.3048 - val_accuracy: 0.9997
Epoch 41/50
844/844 [=====] - 2s 2ms/step - loss: 0.2041 -
accuracy: 0.9065 - val_loss: 0.2796 - val_accuracy: 1.0000
Epoch 42/50
844/844 [=====] - 2s 2ms/step - loss: 0.2041 -
accuracy: 0.9064 - val_loss: 0.3013 - val_accuracy: 1.0000
Epoch 43/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9064 - val_loss: 0.3590 - val_accuracy: 0.9860
Epoch 44/50
844/844 [=====] - 2s 2ms/step - loss: 0.2041 -
accuracy: 0.9064 - val_loss: 0.3004 - val_accuracy: 1.0000
Epoch 45/50
844/844 [=====] - 2s 2ms/step - loss: 0.2038 -

```

```

accuracy: 0.9066 - val_loss: 0.3343 - val_accuracy: 1.0000
Epoch 46/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9065 - val_loss: 0.2917 - val_accuracy: 0.9997
Epoch 47/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9064 - val_loss: 0.2954 - val_accuracy: 1.0000
Epoch 48/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9065 - val_loss: 0.3125 - val_accuracy: 1.0000
Epoch 49/50
844/844 [=====] - 2s 2ms/step - loss: 0.2039 -
accuracy: 0.9065 - val_loss: 0.3090 - val_accuracy: 1.0000
Epoch 50/50
844/844 [=====] - 2s 2ms/step - loss: 0.2040 -
accuracy: 0.9064 - val_loss: 0.3257 - val_accuracy: 0.9960

```

[9]: <tensorflow.python.keras.callbacks.History at 0x7fd9d43b6b50>

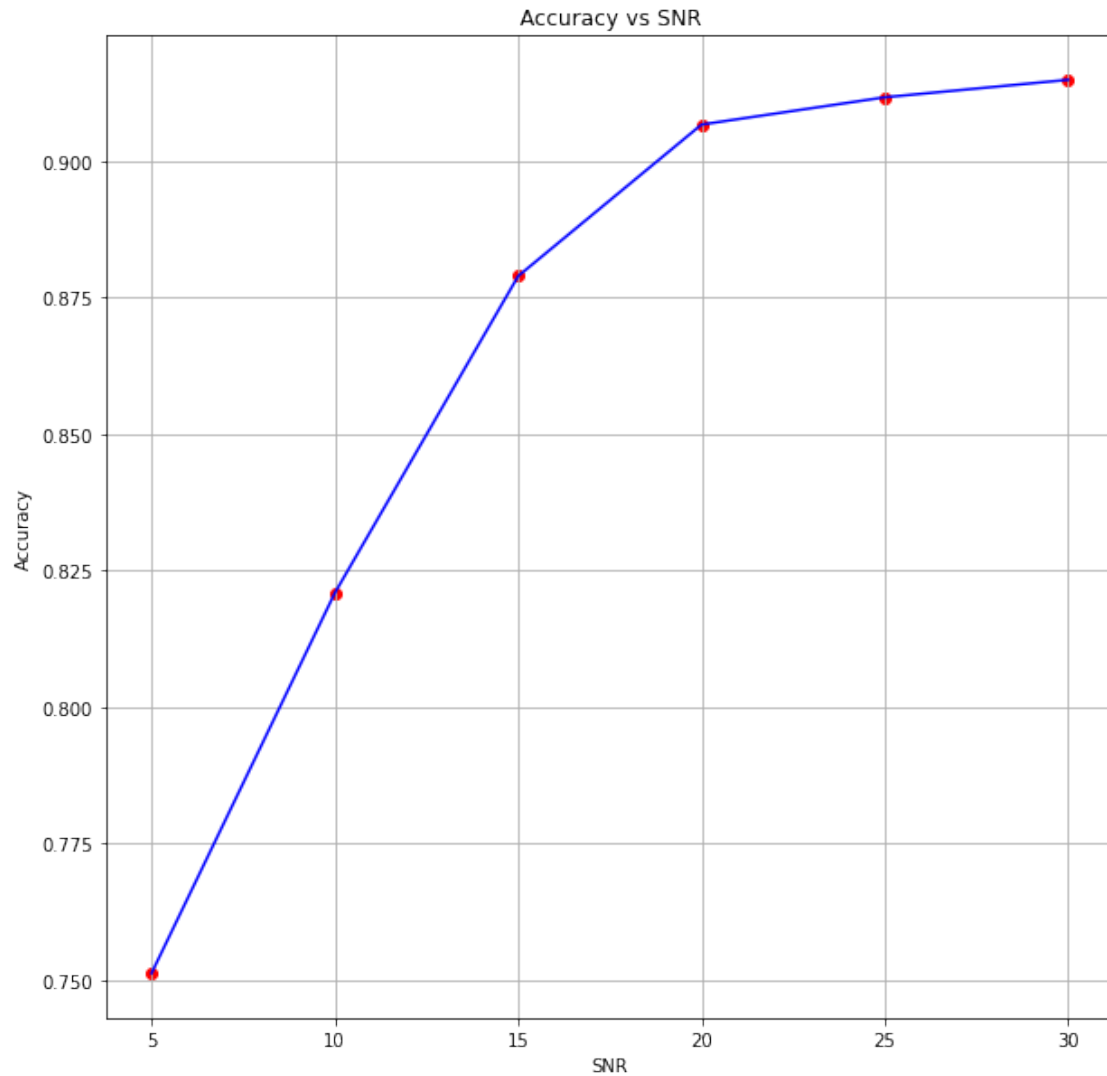
Evaluate the Model

[10]: EvaluateData(AWGN_Model, X_Train, y_Train, X_Valid, y_Valid, "AWGN_Accuracy.
→jpg")

```

Evaluating Model
938/938 [=====] - 2s 2ms/step - loss: 1.7904 -
accuracy: 0.7511
SNR: 5 Accuracy: 0.7511333227157593
938/938 [=====] - 2s 2ms/step - loss: 1.0192 -
accuracy: 0.8208
SNR: 10 Accuracy: 0.8208333253860474
938/938 [=====] - 2s 2ms/step - loss: 0.4816 -
accuracy: 0.8789
SNR: 15 Accuracy: 0.8788999915122986
938/938 [=====] - 2s 2ms/step - loss: 0.2641 -
accuracy: 0.9067
SNR: 20 Accuracy: 0.9066666960716248
938/938 [=====] - 2s 2ms/step - loss: 0.2244 -
accuracy: 0.9117
SNR: 25 Accuracy: 0.9116666913032532
938/938 [=====] - 2s 2ms/step - loss: 0.2170 -
accuracy: 0.9149
SNR: 30 Accuracy: 0.9148666858673096

```



Save Model

```
[11]: AWGN_Model.save("Model/AWGN.h5")
```

1.6 AutoML Model for Rayleigh Channel of Channel-Length = 2

For Rayleigh Channel, Input Dimensions is (100,2)

Rayleigh Data

```
[12]: X_Train, y_Train, X_Valid, y_Valid = GenerateDatasets('Rayleigh',L=2)
```

Creating a CNN Classifier

```
[13]: RayleighClassifier = ak.ImageClassifier(  
    overwrite=True,  
    max_trials=1,  
    loss='categorical_crossentropy',  
    num_classes=3)  
  
RayleighClassifier.fit(X_Train, y_Train, epochs=5, batch_size=16,  
    ↪validation_split=0.3)
```

Trial 1 Complete [00h 00m 16s]

val_loss: 0.11849616467952728

Best val_loss So Far: 0.11849616467952728

Total elapsed time: 00h 00m 16s

INFO:tensorflow:Oracle triggered exit

Epoch 1/5

938/938 [=====] - 3s 3ms/step - loss: 0.1085 -
accuracy: 0.9769

Epoch 2/5

938/938 [=====] - 3s 3ms/step - loss: 0.3147 -
accuracy: 0.8675

Epoch 3/5

938/938 [=====] - 3s 3ms/step - loss: 0.2239 -
accuracy: 0.9413

Epoch 4/5

938/938 [=====] - 3s 3ms/step - loss: 0.8871 -
accuracy: 0.9248

Epoch 5/5

938/938 [=====] - 3s 3ms/step - loss: 0.1001 -
accuracy: 0.9685

INFO:tensorflow:Assets written to: ./image_classifier/best_model/assets

CNN Model Summary

```
[14]: Rayleigh_Model = RayleighClassifier.export_model()  
Rayleigh_Model.summary()  
tf.keras.utils.plot_model(Rayleigh_Model, to_file='Images/Rayleigh_Model_L=2.  
    ↪png', show_shapes=False, show_layer_names=True)
```

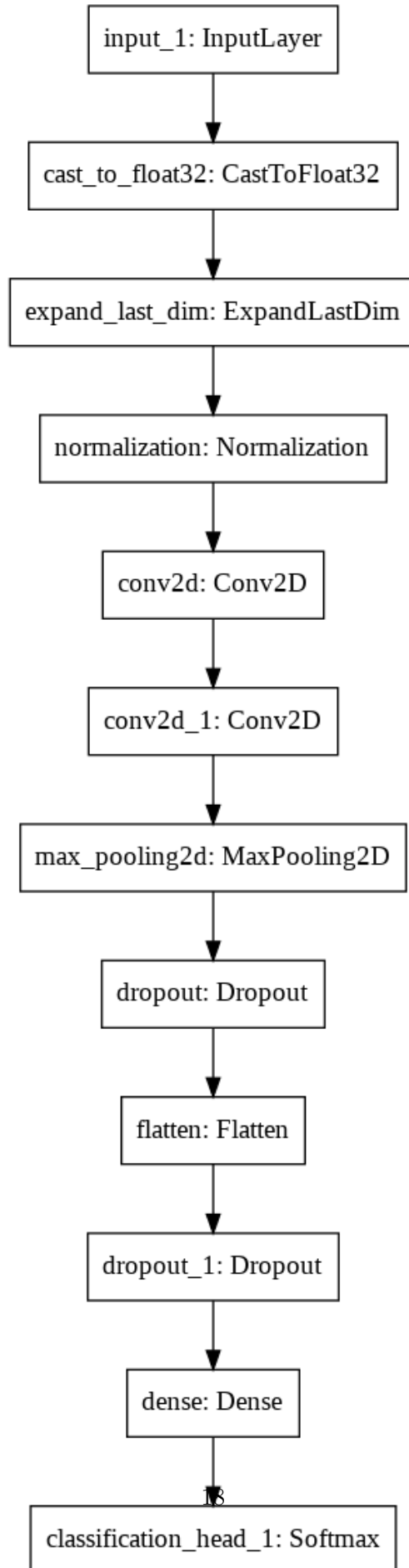
Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 100, 2)]	0
cast_to_float32 (CastToFloat)	(None, 100, 2)	0

expand_last_dim (ExpandLastD	(None, 100, 2, 1)	0
normalization (Normalization	(None, 100, 2, 1)	3
conv2d (Conv2D)	(None, 100, 2, 32)	320
conv2d_1 (Conv2D)	(None, 100, 2, 64)	18496
max_pooling2d (MaxPooling2D)	(None, 50, 1, 64)	0
dropout (Dropout)	(None, 50, 1, 64)	0
flatten (Flatten)	(None, 3200)	0
dropout_1 (Dropout)	(None, 3200)	0
dense (Dense)	(None, 3)	9603
classification_head_1 (Softm	(None, 3)	0

=====
 Total params: 28,422
 Trainable params: 28,419
 Non-trainable params: 3
 =====

[14]:



1.6.1 Training and Evaluating CNN Model

Training the CNN Model

```
[15]: Rayleigh_Model.fit(X_Train, y_Train, epochs=20, batch_size=16,  
    ↪ validation_split=0.3, shuffle=True)
```

Epoch 1/20

657/657 [=====] - 3s 4ms/step - loss: 0.1670 -
accuracy: 0.9650 - val_loss: 0.0312 - val_accuracy: 1.0000

Epoch 2/20

657/657 [=====] - 2s 4ms/step - loss: 8.4203e-04 -
accuracy: 1.0000 - val_loss: 0.0037 - val_accuracy: 1.0000

Epoch 3/20

657/657 [=====] - 2s 4ms/step - loss: 1.9418e-04 -
accuracy: 1.0000 - val_loss: 6.7732e-04 - val_accuracy: 1.0000

Epoch 4/20

657/657 [=====] - 2s 4ms/step - loss: 3.2599e-05 -
accuracy: 1.0000 - val_loss: 1.4871e-04 - val_accuracy: 1.0000

Epoch 5/20

657/657 [=====] - 2s 4ms/step - loss: 9.9750e-06 -
accuracy: 1.0000 - val_loss: 6.0170e-05 - val_accuracy: 1.0000

Epoch 6/20

657/657 [=====] - 2s 4ms/step - loss: 5.3888e-06 -
accuracy: 1.0000 - val_loss: 2.6325e-05 - val_accuracy: 1.0000

Epoch 7/20

657/657 [=====] - 2s 4ms/step - loss: 2.8481e-06 -
accuracy: 1.0000 - val_loss: 1.4253e-05 - val_accuracy: 1.0000

Epoch 8/20

657/657 [=====] - 3s 4ms/step - loss: 1.1521e-06 -
accuracy: 1.0000 - val_loss: 7.4011e-06 - val_accuracy: 1.0000

Epoch 9/20

657/657 [=====] - 3s 4ms/step - loss: 8.6505e-07 -
accuracy: 1.0000 - val_loss: 2.7228e-06 - val_accuracy: 1.0000

Epoch 10/20

657/657 [=====] - 3s 4ms/step - loss: 4.0110e-07 -
accuracy: 1.0000 - val_loss: 1.0809e-06 - val_accuracy: 1.0000

Epoch 11/20

657/657 [=====] - 3s 4ms/step - loss: 2.4870e-07 -
accuracy: 1.0000 - val_loss: 7.2877e-07 - val_accuracy: 1.0000

Epoch 12/20

657/657 [=====] - 3s 4ms/step - loss: 9.6809e-08 -
accuracy: 1.0000 - val_loss: 3.0131e-07 - val_accuracy: 1.0000

Epoch 13/20

657/657 [=====] - 2s 4ms/step - loss: 8.1628e-08 -

```

accuracy: 1.0000 - val_loss: 1.6660e-07 - val_accuracy: 1.0000
Epoch 14/20
657/657 [=====] - 2s 4ms/step - loss: 6.5700e-08 -
accuracy: 1.0000 - val_loss: 1.1648e-07 - val_accuracy: 1.0000
Epoch 15/20
657/657 [=====] - 2s 4ms/step - loss: 3.1255e-08 -
accuracy: 1.0000 - val_loss: 8.5513e-08 - val_accuracy: 1.0000
Epoch 16/20
657/657 [=====] - 2s 4ms/step - loss: 2.6317e-08 -
accuracy: 1.0000 - val_loss: 6.7976e-08 - val_accuracy: 1.0000
Epoch 17/20
657/657 [=====] - 3s 4ms/step - loss: 4.7647e-08 -
accuracy: 1.0000 - val_loss: 3.0973e-07 - val_accuracy: 1.0000
Epoch 18/20
657/657 [=====] - 2s 4ms/step - loss: 2.8417e-08 -
accuracy: 1.0000 - val_loss: 1.2347e-07 - val_accuracy: 1.0000
Epoch 19/20
657/657 [=====] - 2s 4ms/step - loss: 9.6162e-09 -
accuracy: 1.0000 - val_loss: 2.1723e-09 - val_accuracy: 1.0000
Epoch 20/20
657/657 [=====] - 2s 4ms/step - loss: 1.5395e-08 -
accuracy: 1.0000 - val_loss: 7.9473e-11 - val_accuracy: 1.0000

```

[15]: <tensorflow.python.keras.callbacks.History at 0x7fd9d464eb10>

Evaluate the CNN Model

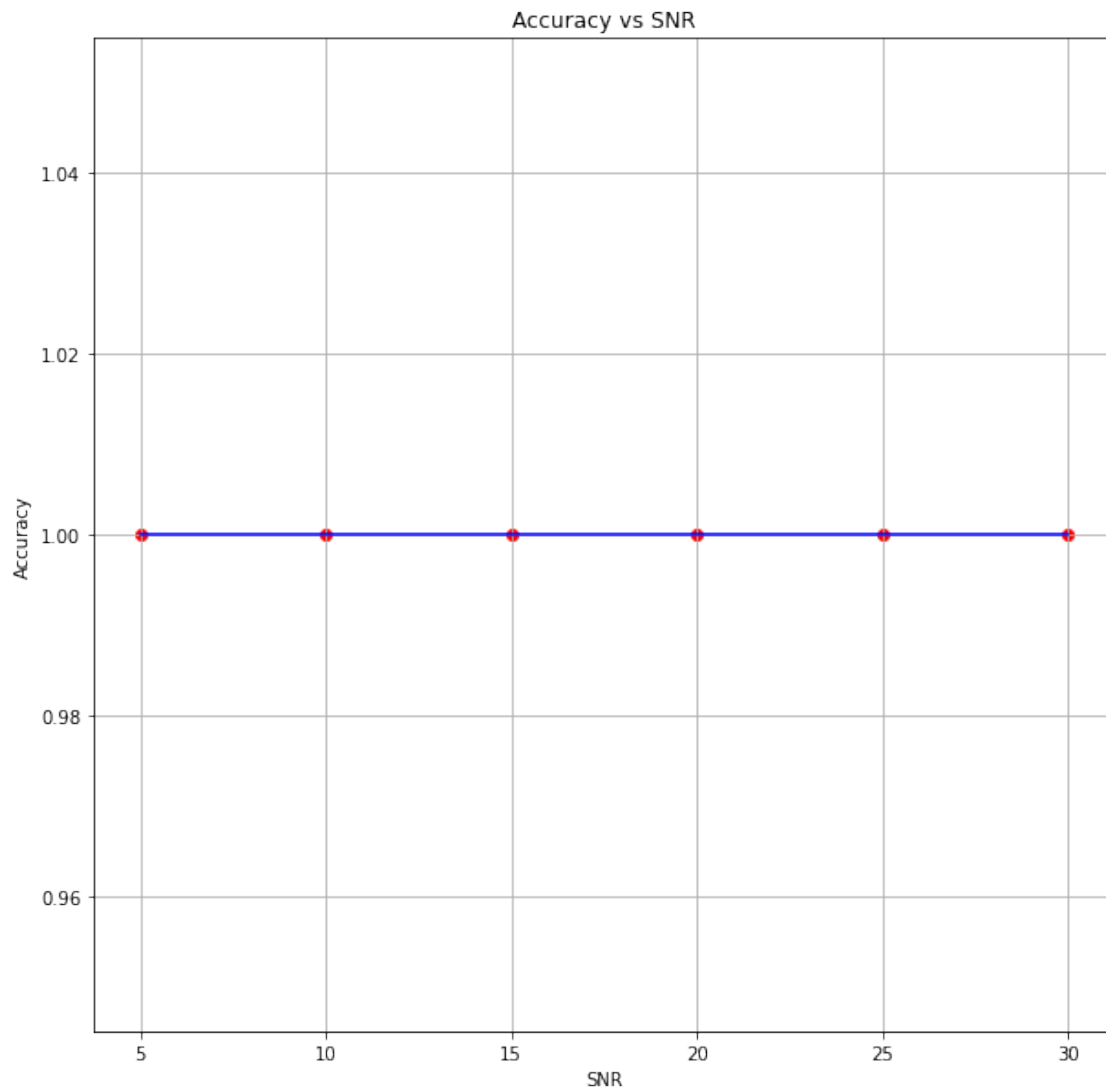
[16]: EvaluateData(Rayleigh_Model, X_Train, y_Train, X_Valid, y_Valid,
 ↪ "Rayleigh_Accuracy_L=2.jpg")

```

Evaluating Model
469/469 [=====] - 1s 2ms/step - loss: 5.7395e-08 -
accuracy: 1.0000
SNR: 5 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 2.6385e-09 -
accuracy: 1.0000
SNR: 10 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 3.2584e-10 -
accuracy: 1.0000
SNR: 15 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 1.0331e-10 -
accuracy: 1.0000
SNR: 20 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 1.9868e-10 -
accuracy: 1.0000
SNR: 25 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 1.2716e-10 -
accuracy: 1.0000

```

SNR: 30 Accuracy: 1.0



Save CNN Model

```
[17]: Rayleigh_Model.save("Model/Rayleigh_L=2.h5")
```

1.7 AutoML Model for Rayleigh Channel of Channel-Length = 3

For Rayleigh Channel, Input Dimensions is (100,2)

Rayleigh Data

```
[18]: X_Train, y_Train, X_Valid, y_Valid = GenerateDatasets('Rayleigh',L=3)
```

Creating a CNN Classifier

```
[19]: RayleighClassifier = ak.ImageClassifier(  
      overwrite=True,  
      max_trials=1,  
      loss='categorical_crossentropy',  
      num_classes=3)  
  
RayleighClassifier.fit(X_Train, y_Train, epochs=5, batch_size=16,  
      ↪validation_split=0.3)
```

Trial 1 Complete [00h 00m 14s]

val_loss: 0.1409987360239029

Best val_loss So Far: 0.1409987360239029

Total elapsed time: 00h 00m 14s

INFO:tensorflow:Oracle triggered exit

Epoch 1/5

938/938 [=====] - 3s 3ms/step - loss: 0.0913 -
accuracy: 0.9736

Epoch 2/5

938/938 [=====] - 3s 3ms/step - loss: 0.8848 -
accuracy: 0.9169

Epoch 3/5

938/938 [=====] - 3s 3ms/step - loss: 0.4379 -
accuracy: 0.9280

Epoch 4/5

938/938 [=====] - 3s 3ms/step - loss: 0.3910 -
accuracy: 0.9111

Epoch 5/5

938/938 [=====] - 3s 3ms/step - loss: 0.3136 -
accuracy: 0.9426

INFO:tensorflow:Assets written to: ./image_classifier/best_model/assets

CNN Model Summary

```
[20]: Rayleigh_Model = RayleighClassifier.export_model()  
Rayleigh_Model.summary()  
tf.keras.utils.plot_model(Rayleigh_Model, to_file='Images/Rayleigh_Model_L=3.  
    ↪png', show_shapes=False, show_layer_names=True)
```

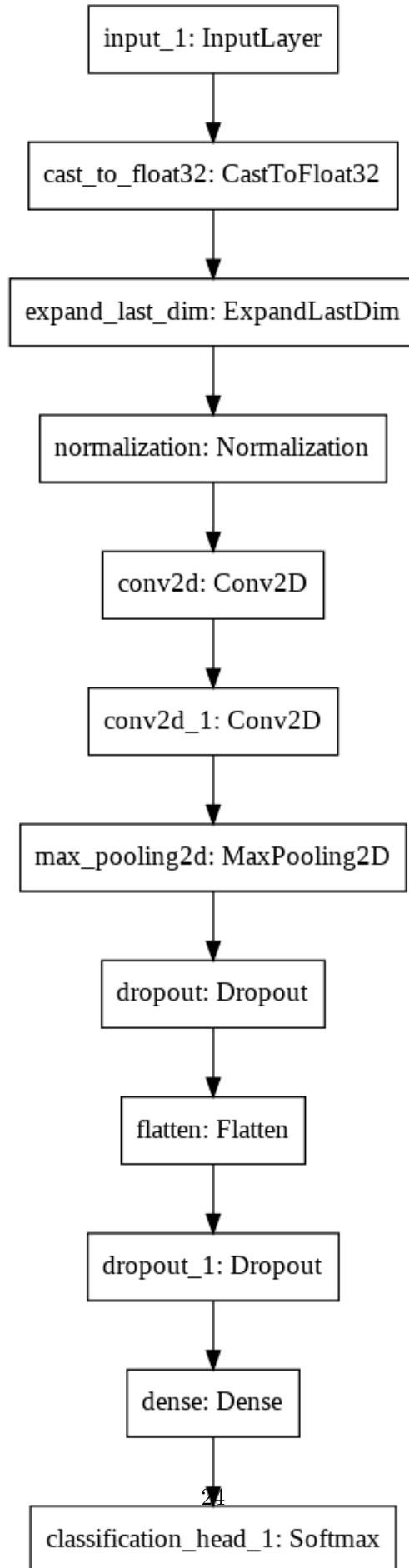
Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 100, 2)]	0
cast_to_float32 (CastToFloat)	(None, 100, 2)	0

expand_last_dim (ExpandLastD	(None, 100, 2, 1)	0
normalization (Normalization	(None, 100, 2, 1)	3
conv2d (Conv2D)	(None, 100, 2, 32)	320
conv2d_1 (Conv2D)	(None, 100, 2, 64)	18496
max_pooling2d (MaxPooling2D)	(None, 50, 1, 64)	0
dropout (Dropout)	(None, 50, 1, 64)	0
flatten (Flatten)	(None, 3200)	0
dropout_1 (Dropout)	(None, 3200)	0
dense (Dense)	(None, 3)	9603
classification_head_1 (Softm	(None, 3)	0

=====
 Total params: 28,422
 Trainable params: 28,419
 Non-trainable params: 3
 =====

[20]:



1.7.1 Training and Evaluating CNN Model

Training the CNN Model

```
[21]: Rayleigh_Model.fit(X_Train, y_Train, epochs=20, batch_size=16,  
    ↪ validation_split=0.3, shuffle=True)
```

Epoch 1/20

657/657 [=====] - 3s 4ms/step - loss: 0.1795 -
accuracy: 0.9681 - val_loss: 0.0129 - val_accuracy: 1.0000

Epoch 2/20

657/657 [=====] - 2s 4ms/step - loss: 5.3318e-04 -
accuracy: 1.0000 - val_loss: 0.0014 - val_accuracy: 1.0000

Epoch 3/20

657/657 [=====] - 2s 4ms/step - loss: 8.8514e-05 -
accuracy: 1.0000 - val_loss: 3.9895e-04 - val_accuracy: 1.0000

Epoch 4/20

657/657 [=====] - 3s 4ms/step - loss: 3.5372e-05 -
accuracy: 1.0000 - val_loss: 1.4485e-04 - val_accuracy: 1.0000

Epoch 5/20

657/657 [=====] - 2s 4ms/step - loss: 1.6374e-05 -
accuracy: 1.0000 - val_loss: 7.3171e-05 - val_accuracy: 1.0000

Epoch 6/20

657/657 [=====] - 2s 4ms/step - loss: 7.3211e-06 -
accuracy: 1.0000 - val_loss: 3.4351e-05 - val_accuracy: 1.0000

Epoch 7/20

657/657 [=====] - 2s 4ms/step - loss: 5.0201e-06 -
accuracy: 1.0000 - val_loss: 1.9795e-05 - val_accuracy: 1.0000

Epoch 8/20

657/657 [=====] - 2s 4ms/step - loss: 3.7358e-06 -
accuracy: 1.0000 - val_loss: 1.1241e-05 - val_accuracy: 1.0000

Epoch 9/20

657/657 [=====] - 2s 4ms/step - loss: 2.0092e-06 -
accuracy: 1.0000 - val_loss: 6.8868e-06 - val_accuracy: 1.0000

Epoch 10/20

657/657 [=====] - 2s 4ms/step - loss: 1.4446e-06 -
accuracy: 1.0000 - val_loss: 3.8716e-06 - val_accuracy: 1.0000

Epoch 11/20

657/657 [=====] - 2s 4ms/step - loss: 7.6345e-07 -
accuracy: 1.0000 - val_loss: 2.5221e-06 - val_accuracy: 1.0000

Epoch 12/20

657/657 [=====] - 2s 4ms/step - loss: 8.3578e-07 -
accuracy: 1.0000 - val_loss: 1.5626e-06 - val_accuracy: 1.0000

Epoch 13/20

657/657 [=====] - 2s 4ms/step - loss: 3.8577e-07 -

```

accuracy: 1.0000 - val_loss: 9.9079e-07 - val_accuracy: 1.0000
Epoch 14/20
657/657 [=====] - 3s 4ms/step - loss: 2.4486e-07 -
accuracy: 1.0000 - val_loss: 6.0670e-07 - val_accuracy: 1.0000
Epoch 15/20
657/657 [=====] - 2s 4ms/step - loss: 2.1856e-07 -
accuracy: 1.0000 - val_loss: 4.6168e-07 - val_accuracy: 1.0000
Epoch 16/20
657/657 [=====] - 2s 4ms/step - loss: 1.4611e-07 -
accuracy: 1.0000 - val_loss: 2.3855e-07 - val_accuracy: 1.0000
Epoch 17/20
657/657 [=====] - 2s 4ms/step - loss: 1.4674e-07 -
accuracy: 1.0000 - val_loss: 2.4160e-07 - val_accuracy: 1.0000
Epoch 18/20
657/657 [=====] - 2s 4ms/step - loss: 1.0978e-07 -
accuracy: 1.0000 - val_loss: 1.1921e-07 - val_accuracy: 1.0000
Epoch 19/20
657/657 [=====] - 2s 4ms/step - loss: 3.9614e-07 -
accuracy: 1.0000 - val_loss: 3.2425e-07 - val_accuracy: 1.0000
Epoch 20/20
657/657 [=====] - 2s 4ms/step - loss: 7.2274e-08 -
accuracy: 1.0000 - val_loss: 1.0724e-07 - val_accuracy: 1.0000

```

[21]: <tensorflow.python.keras.callbacks.History at 0x7fda2011c110>

Evaluate the CNN Model

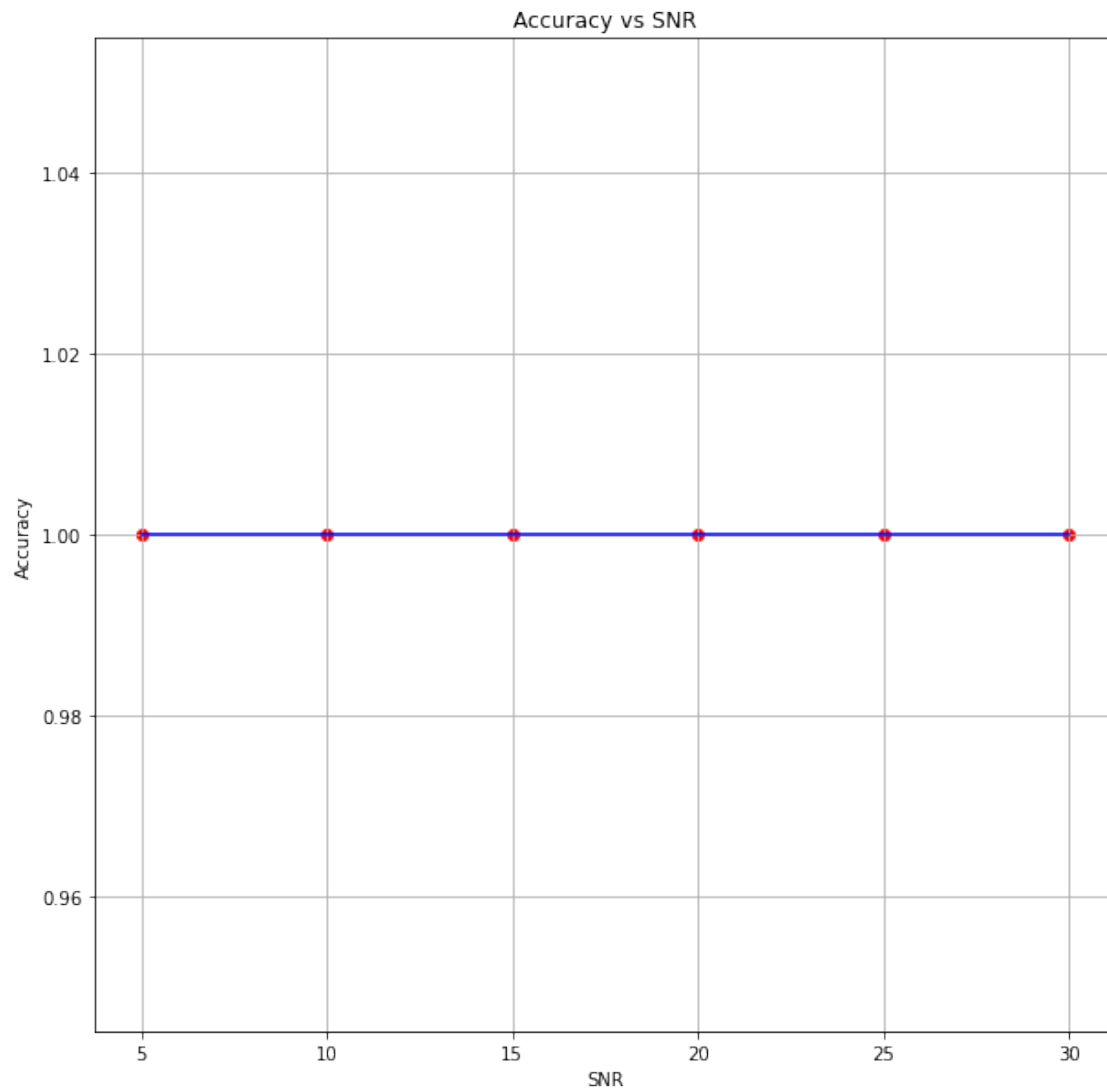
[22]: EvaluateData(Rayleigh_Model, X_Train, y_Train, X_Valid, y_Valid,
↳ "Rayleigh_Accuracy_L=3.jpg")

```

Evaluating Model
469/469 [=====] - 1s 2ms/step - loss: 4.9734e-08 -
accuracy: 1.0000
SNR: 5 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 3.9212e-08 -
accuracy: 1.0000
SNR: 10 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 3.7503e-08 -
accuracy: 1.0000
SNR: 15 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 3.6589e-08 -
accuracy: 1.0000
SNR: 20 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 3.5834e-08 -
accuracy: 1.0000
SNR: 25 Accuracy: 1.0
469/469 [=====] - 1s 2ms/step - loss: 3.5779e-08 -
accuracy: 1.0000

```

SNR: 30 Accuracy: 1.0



Save CNN Model

```
[23]: Rayleigh_Model.save("Model/Rayleigh_L=3.h5")
```