

CAP 6619 - Deep Learning

Project: Genomic Benchmarking

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[Colab Project Link](#)

[Project Github Link](#)

PIP Package Installation, Setup, and Imports

```
In [51]: %%capture_code --path "/content/code/pip_install.py"
# PIP Package Installation
!pip install -q mermaid-py
!pip install -q genomic-benchmarks
!pip install -q jupyter_capture_output
```

Output saved by creating file at /content/code/pip_install.py.

```
In [50]: %%capture_code --path "/content/code/setup_and_imports.py"
# Setup and Imports
from pathlib import Path
import random
import os
import warnings

from genomic_benchmarks.data_check import (
    info,
    is_downloaded,
    list_datasets,
)
from genomic_benchmarks.loc2seq import download_dataset
import jupyter_capture_output
import keras
from keras import backend as K
from keras.layers import (
    Activation,
    BatchNormalization,
    Conv1D,
    Dense,
    Dropout,
    Flatten,
    GlobalAveragePooling1D,
    Input,
    Lambda,
    MaxPooling1D,
    TextVectorization,
)
from keras.losses import (
    BinaryCrossentropy,
    CategoricalCrossentropy,
)
from keras.metrics import (
    BinaryAccuracy,
    CategoricalAccuracy,
```

```

)
from keras.models import Sequential
import keras.ops as ops
import matplotlib.pyplot as plt
from mermaid import Mermaid
import numpy as np
import pandas as pd
import tensorflow as tf

SEED = 1234

os.environ["PYTHONHASHSEED"] = str(SEED)
random.seed(SEED)
np.random.seed(SEED)
keras.utils.set_random_seed(SEED)
tf.random.set_seed(SEED)

# Suppress Keras warnings
warnings.filterwarnings("ignore")

```

Note: The /content/code directory was successfully created.
Output saved by creating file at /content/code/setup_and_imports.py.

Datasets

In [52]:

```
%%capture_code --path "/content/code/datasets.py"
from genomic_benchmarks.data_check import list_datasets

list_datasets()
```

Out[52]:

```
['demo_human_or_worm',
'drosophila_enhancers_stark',
'dummy_mouse_enhancers_ensembl',
'human_ensembl_regulatory',
'human_enhancers_cohn',
'demo_coding_vs_intergenic_seqs',
'human_nontata_promoters',
'human_enhancers_ensembl',
'human_ocr_ensembl']
```

Output saved by creating file at /content/code/datasets.py.

In [53]:

```
%%capture_code --path "/content/code/download_datasets.py"
selected_dataset_list = [
    "dummy_mouse_enhancers_ensembl",
    "drosophila_enhancers_stark",
    "human_enhancers_cohn",
    "human_nontata_promoters",
]

for dataset in selected_dataset_list:
    if not is_downloaded(dataset):
        download_dataset(dataset)
    print(info(dataset, description=True))
    print("\n")
```

Output saved by creating file at /content/code/download_datasets.py.

Create Training and Test Sets

```
In [6]: batch_size = 64
selected_dataset = selected_dataset_list[1]
SEQ_PATH = Path.home() / ".genomic_benchmarks" / selected_dataset
SEQ_PATH
```

```
Out[6]: PosixPath('/root/.genomic_benchmarks/drosophila_enancers_stark')
```

```
In [7]: classes = [
    x.stem for x
    in (SEQ_PATH/"train").iterdir()
    if x.is_dir()
]
num_classes = len(classes)

train_set = tf.keras.preprocessing.text_dataset_from_directory(
    SEQ_PATH / "train",
    batch_size=batch_size,
    class_names=classes,
    shuffle=True,
    seed=SEED,
)

test_set = tf.keras.preprocessing.text_dataset_from_directory(
    SEQ_PATH / "test",
    batch_size=batch_size,
    class_names=classes,
)

if num_classes > 2:
    train_set = train_set.map(
        lambda x, y: (x, tf.one_hot(y, depth=num_classes)))

if num_classes > 2:
    test_set = test_set.map(
        lambda x, y: (x, tf.one_hot(y, depth=num_classes)))
```

Found 5184 files belonging to 2 classes.

Found 1730 files belonging to 2 classes.

```
In [8]: print(f"Training Set Type: {type(train_set)}")
print(f"Testing Set Type: {type(test_set)}")
```

Training Set Type: <class 'tensorflow.python.data.ops.prefetch_op._PrefetchDataset'>
Testing Set Type: <class 'tensorflow.python.data.ops.prefetch_op._PrefetchDataset'>

Vectorize the Dataset

NOTE: This code for the model and vectorize layer had to be update for use with python3.11 and Keras>=3.0

```
In [54]: %%capture_code --path "/content/code/vectorize_layer.py"
char_split_fn = lambda x: tf.strings.unicode_split(x, input_encoding="UTF-8")
vectorize_layer = keras.layers.TextVectorization(
    output_mode="int",
    split=char_split_fn,
)

vectorize_layer.adapt(train_set.map(lambda x, y: x))
vocab_size = len(vectorize_layer.get_vocabulary())
vectorize_layer.get_vocabulary()
```

```
Output saved by creating file at /content/code/vectorize_layer.py.  
Out[54]: ['', '[UNK]', np.str_('t'), np.str_('a'), np.str_('c'), np.str_('g')]
```

```
In [55]: %%capture_code --path "/content/code/vectorize_text.py"  
def vectorize_text(text, label):  
    """Returns a vector representation of the text.  
  
    :param text: The text to vectorize  
    :param label: The label of the text  
    :return: A vector representation of the text  
    """  
    text = tf.expand_dims(text, axis=-1)  
    return vectorize_layer(text)-2, label
```

```
Output saved by creating file at /content/code/vectorize_text.py.
```

```
In [11]: train_ds = train_set.map(vectorize_text)  
test_ds = test_set.map(vectorize_text)  
  
print(f"Training Set Type: {type(train_ds)}")  
print(f"Testing Set Type: {type(test_ds)}")
```

```
Training Set Type: <class 'tensorflow.python.data.ops.map_op._MapDataset'>  
Testing Set Type: <class 'tensorflow.python.data.ops.map_op._MapDataset'>
```

Check Sequence Length

```
In [56]: %%capture_code --path "/content/code/check_lengths.py"  
def check_seq_lengths(dataset, use_padding):  
    """Returns the maximum sequence length and the length of the sequence with  
    tokens.  
  
    :param dataset: List of sequences  
    :param use_padding: Padding  
    :return: Maximum sequence length and length of sequence with tokens  
    """  
    max_seq_len = max([len(dataset[i][0]) for i in range(len(dataset))])  
    print(f"Max Sequence Length: {max_seq_len}")  
    same_length = [len(dataset[i][0]) == max_seq_len for i in range(len(dataset))]  
    if not all(same_length):  
        print("not all sequences are of the same length")  
  
    if use_padding:  
        len_with_tokens = max_seq_len + 3  
    else:  
        len_with_tokens = max_seq_len + 2  
  
    return max_seq_len, len_with_tokens
```

```
Output saved by creating file at /content/code/check_lengths.py.
```

```
In [13]: data_list = list(train_ds.as_numpy_iterator())  
max_seq_len, nn_input_len = check_seq_lengths(data_list, use_padding=True)
```

```
Max Sequence Length: 64
```

Update Dataset

```
In [57]: %%capture_code --path "/content/code/update_train_test_sets.py"  
def update_train_test_sets(id_num, batch_size=64):  
    """Updates the training and testing sets, and returns the name of the  
    dataset.
```

```

:param id_num: The ID number of the dataset
:param batch_size: The batch size
"""
batch_size = batch_size
selected_dataset = selected_dataset_list[id_num]
SEQ_PATH = Path.home() / ".genomic_benchmarks" / selected_dataset

classes = [
    x.stem for x
    in (SEQ_PATH/"train").iterdir()
    if x.is_dir()
]
num_classes = len(classes)

train_set = tf.keras.preprocessing.text_dataset_from_directory(
    SEQ_PATH / "train",
    batch_size=batch_size,
    class_names=classes,
    shuffle=True,
    seed=SEED,
)

test_set = tf.keras.preprocessing.text_dataset_from_directory(
    SEQ_PATH / "test",
    batch_size=batch_size,
    class_names=classes,
)

if num_classes > 2:
    train_set = train_set.map(
        lambda x, y: (x, tf.one_hot(y, depth=num_classes)))

if num_classes > 2:
    test_set = test_set.map(
        lambda x, y: (x, tf.one_hot(y, depth=num_classes)))

vectorize_layer.adapt(train_set.map(lambda x, y: x))
vocab_size = len(vectorize_layer.get_vocabulary())
vectorize_layer.get_vocabulary()

train_ds = train_set.map(vectorize_text)
test_ds = test_set.map(vectorize_text)

return selected_dataset

```

Output saved by creating file at /content/code/update_train_test_sets.py.

Model Creation

Metrics

```

In [58]: %%capture_code --path "/content/code/f1_score.py"
def f1_score(y_true, y_pred):
    """Returns the F1 score.

:param y_true: The true labels
:param y_pred: The predicted labels
:return: The F1 score
"""

def precision(y_true, y_pred):

```

```

"""Returns the precision.

:param y_true: The true labels
:param y_pred: The predicted labels
:return: The precision
"""

true_positives = ops.sum(
    ops.round(ops.clip(tf.cast(y_true, tf.float32) * y_pred, 0, 1)))
predicted_positives = ops.sum(ops.round(ops.clip(y_pred, 0, 1)))
precision = true_positives / (predicted_positives + K.epsilon())
return precision

def recall(y_true, y_pred):
    """Returns the recall.

    :param y_true: The true labels
    :param y_pred: The predicted labels
    :return: The recall
    """

    true_positives = ops.sum(
        ops.round(ops.clip(tf.cast(y_true, tf.float32) * y_pred, 0, 1)))
    possible_positives = tf.cast(
        ops.sum(ops.round(ops.clip(y_true, 0, 1))), tf.float32)
    recall = (
        tf.cast(true_positives, tf.float32)
        / (possible_positives + K.epsilon()))
    )
    return recall

precision = precision(y_true, y_pred)
recall = recall(y_true, y_pred)
return 2 * ((precision * recall) / (precision + recall + K.epsilon()))

```

Output saved by creating file at /content/code/f1_score.py.

Initial Model

```

In [64]: %%capture_code --path "/content/code/initial_model.py"
def create_basic_cnn_model(num_classes, vocab_size):
    """Returns a basic CNN model and model name.

    :param num_classes: The number of classes to classify
    :param vocab_size: The size of the vocabulary
    :return: Model name and CNN model
    """

    name = "basic"

    if num_classes == 2:
        last_layer = Dense(1, activation="sigmoid")
        loss = BinaryCrossentropy(from_logits=True)
        acc = BinaryAccuracy()

    else:
        last_layer = Dense(num_classes, activation="softmax")
        loss = "categorical_crossentropy"
        acc = CategoricalAccuracy()

    onehot_layer = Lambda(
        lambda x: tf.one_hot(tf.cast(x, "int64"), depth=vocab_size))

    print(onehot_layer)

```

```

model = Sequential(
    [
        onehot_layer,
        Conv1D(
            filters=32,
            kernel_size=8,
            data_format="channels_last",
            activation="relu",
        ),
        BatchNormalization(),
        MaxPooling1D(),
        Conv1D(
            filters=16,
            kernel_size=8,
            data_format="channels_last",
            activation="relu",
        ),
        BatchNormalization(),
        MaxPooling1D(),
        Conv1D(
            filters=4,
            kernel_size=8,
            data_format="channels_last",
            activation="relu",
        ),
        BatchNormalization(),
        MaxPooling1D(),
        Dropout(0.3),
        GlobalAveragePooling1D(),
        last_layer,
    ]
)

model.compile(
    optimizer="adam",
    loss=loss,
    metrics=[acc, f1_score],
)

```

return name, model

Output saved by creating file at /content/code/initial_model.py.

Final Model

NOTE: This model has the updated architecture adding the Global Average Pooling Layer

```

In [62]: %%capture_code --path "/content/code/final_model.py"
def create_final_cnn_model(num_classes, vocab_size):
    """Returns a CNN model and model name.

    :param num_classes: The number of classes to classify
    :param vocab_size: The size of the vocabulary
    :return: Model name and CNN model
    """
    name = "final"

    if num_classes == 2:
        last_layer = Dense(1, activation="sigmoid")
        loss = BinaryCrossentropy(from_logits=True)
        acc = BinaryAccuracy(threshold=0.5)
    else:

```

```

last_layer = Dense(num_classes, activation="softmax")
loss = "categorical_crossentropy"
acc = CategoricalAccuracy()

onehot_layer = Lambda(
    lambda x: tf.one_hot(tf.cast(x, "int64"), depth=vocab_size))

model = Sequential(
    [
        onehot_layer,
        Conv1D(
            filters=16,
            kernel_size=8,
            data_format="channels_last",
        ),
        BatchNormalization(),
        Activation("relu"),
        MaxPooling1D(),
        Conv1D(
            filters=8,
            kernel_size=8,
            data_format="channels_last",
            activation="relu",
        ),
        BatchNormalization(),
        MaxPooling1D(),
        Conv1D(
            filters=4,
            kernel_size=8,
            data_format="channels_last",
            activation="relu",
        ),
        BatchNormalization(),
        MaxPooling1D(),
        GlobalAveragePooling1D(),
        Flatten(),
        Dense(units=512, activation="relu"),
        last_layer,
    ]
)

model.compile(
    optimizer="adam",
    loss=loss,
    metrics=[acc, f1_score],
)

```

return name, model

Output saved by creating file at /content/code/final_model.py.

Run Models

Metrics and Plots

```
In [67]: %%capture_code --path "/content/code/display_metrics.py"
def plot_metrics(model_info_dict, set_name, horizontal=False):
```

```

"""Shows the plots for the training accuracy, f1 score, and loss.
Then shows the scores for the model.

:param model_info_dict: Dictionary of model information
:param set_name: Name of dataset
:param horizontal: Whether to show the plots horizontally
"""
b_hist = model_info_dict["basic_history"]
acc_1 = np.array(b_hist.history["binary_accuracy"])
f1_1 = np.array(b_hist.history["f1_score"])
loss_1 = np.array(b_hist.history["loss"])

f_hist = model_info_dict["final_history"]
acc_2 = np.array(f_hist.history["binary_accuracy"])
f1_2 = np.array(f_hist.history["f1_score"])
loss_2 = np.array(f_hist.history["loss"])

# Shifted the starting index to start at 1 instead of 0
epochs = np.arange(loss_1.shape[0]) + 1

fig_size = (15, 5) if horizontal else (5, 15)
plt.figure(figsize=fig_size)
models = [model_info_dict["basic_name"], model_info_dict["final_name"]]

rows = 1 if horizontal else 3
cols = 3 if horizontal else 1

plt.subplot(rows, cols, 1)
plt.plot(epochs, acc_1, epochs, acc_2)
plt.title("Training Accuracy")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend(models, loc="lower right")

plt.subplot(rows, cols, 2)
plt.plot(epochs, f1_1, epochs, f1_2, linestyle="--")
plt.title("Training F1 Score")
plt.xlabel("Epochs")
plt.ylabel("")
plt.legend(models, loc="lower right")

plt.subplot(rows, cols, 3)
plt.plot(epochs, loss_1, epochs, loss_2)
plt.title("Loss")
plt.xlabel("Epochs")
plt.ylabel("Crossentropy Loss")
plt.legend(models, loc="upper right")

plt.show()

for m in models:
    eval = model_info_dict[f"{m}_evaluation"]
    print(f"{m.title()} Model for Set: {set_name}:")
    print(f"Total Loss: {round(eval['loss'], 6)}")
    print(f"Accuracy: {round(eval['binary_accuracy'] * 100, 2)}%")
    print(f"F1 Score: {round(eval['f1_score'], 2)}")
    print()

```

Output saved by creating file at /content/code/display_metrics.py.

Train and Evaluate Models

```
In [68]: %%capture_code --path "/content/code/training_model.py"
def train_and_evaluate_models(set_name, epochs):
    """Returns the training and evaluation the metrics for the models.

    :param set_name: Name of dataset
    :param epochs: Number of epochs to train
    :return: Dictionary of model information
    """
    model_info = {
        "basic_name": None,
        "basic_history": None,
        "basic_evaluation": None,
        "final_name": None,
        "final_history": None,
        "final_evaluation": None,
    }

    print(f"Training Models for: {set_name}")
    basic_name, basic_model = create_basic_cnn_model(num_classes, vocab_size)
    model_info["basic_name"] = basic_name
    basic_history = basic_model.fit(
        train_ds,
        epochs=epochs,
        verbose=0,
    )
    model_info["basic_history"] = basic_history
    model_info["basic_evaluation"] = basic_model.evaluate(
        test_ds, verbose=0, return_dict=True)

    final_name, final_model = create_final_cnn_model(num_classes, vocab_size)
    model_info["final_name"] = final_name
    final_history = final_model.fit(
        train_ds,
        epochs=epochs,
        verbose=0,
    )
    model_info["final_history"] = final_history
    model_info["final_evaluation"] = final_model.evaluate(
        test_ds, verbose=0, return_dict=True)

    return model_info
```

Output saved by creating file at /content/code/training_model.py.

```
In [20]: epochs = 10
```

Mouse Enhancer regions vs Random background regions from GRCm38

Dataset 0: dummy_mouse_enancers_ensemble

- The length of genomic intervals ranges from 331 to 4776, with average 2369.5769 and median 2381.0
- Totally 1210 sequences have been found, 968 for training and 242 for testing

```
In [72]: dataset = update_train_test_sets(0)
model_0 = train_and_evaluate_models(dataset, epochs)
```

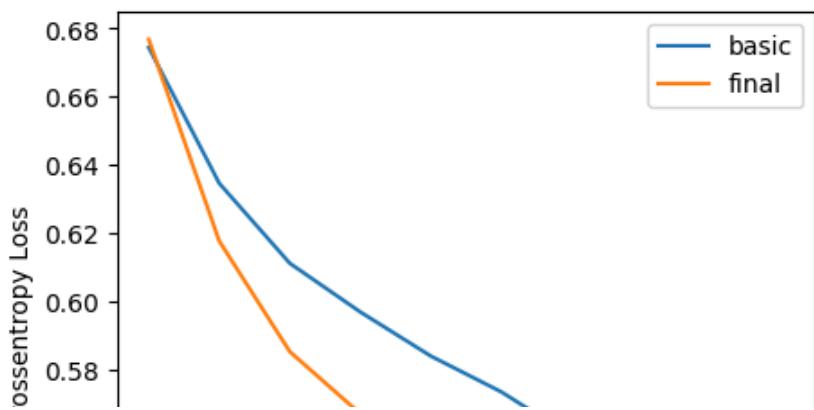
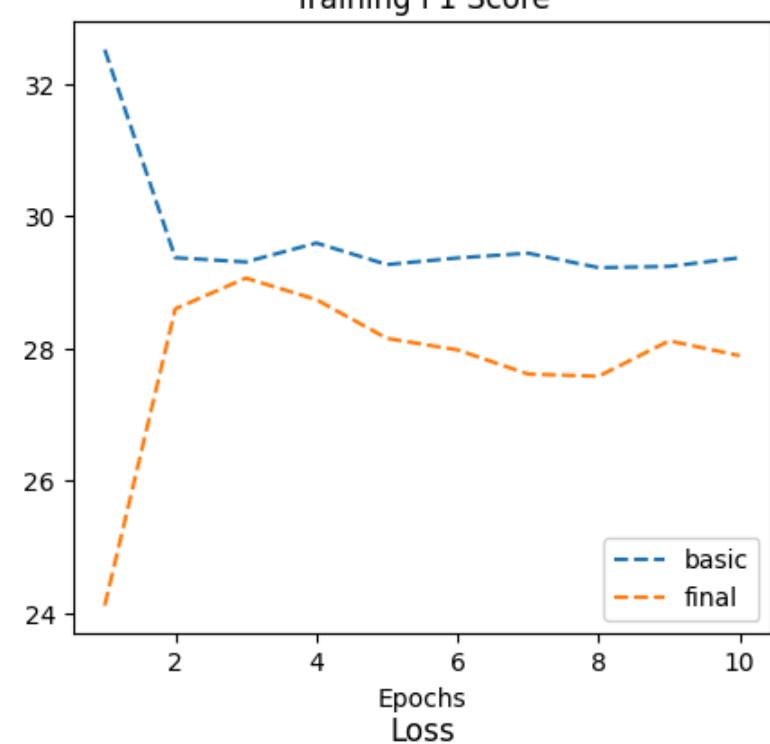
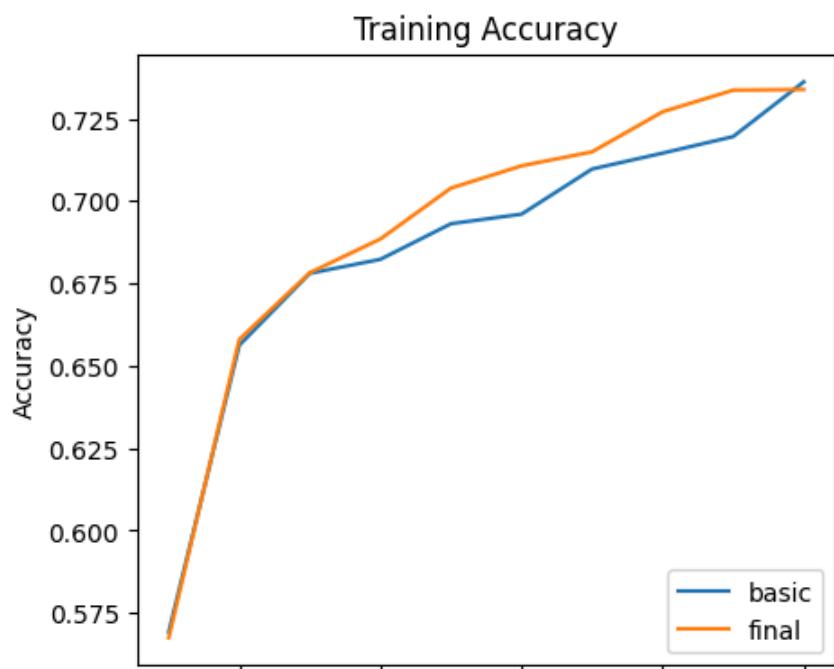
Output saved by overwring previous file at /content/code/example_model_run.py.

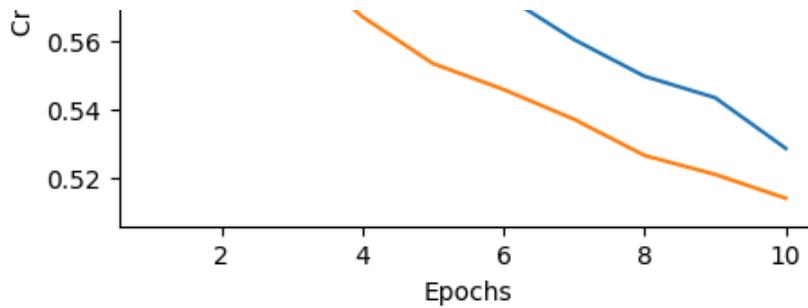
```
In [22]: %%capture_img --path "/content/images/model_0.png"
%%capture_text --path "/content/images/model_0.txt"
```

```
plot_metrics(model_0, dataset)
```

Note: The /content/images directory was successfully created.
Output saved by creating file at /content/images/model_0.png.
Output saved by creating file at /content/images/model_0.txt.
Basic Model for Set: dummy_mouse_enhancers_ensembl:
Total Loss: 0.579737
Accuracy: 66.82%
F1 Score: 33.13

Final Model for Set: dummy_mouse_enhancers_ensembl:
Total Loss: 0.999781
Accuracy: 50.64%
F1 Score: 0.76



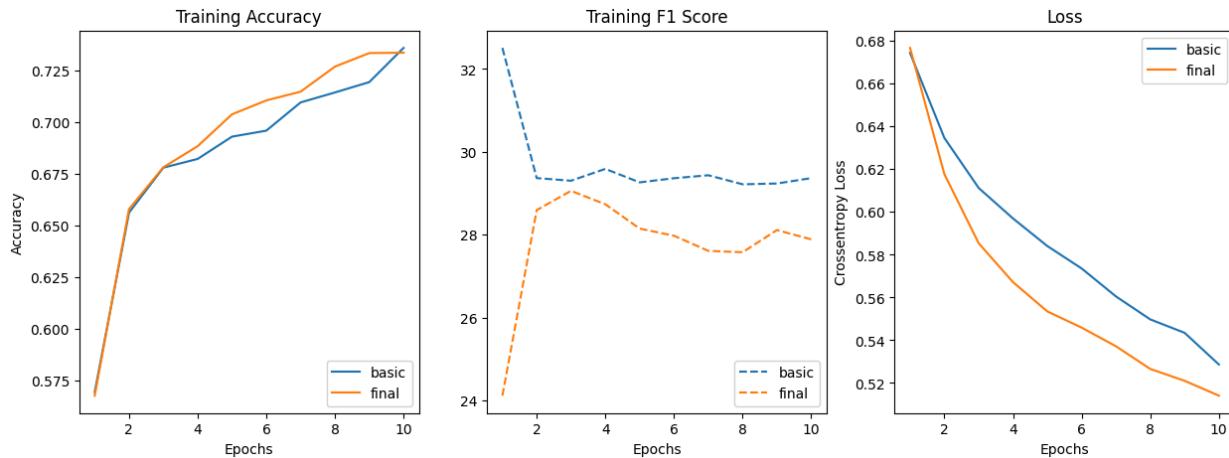


In [23]: # For presentation

```
%capture_img --path "/content/images/model_0_pp.png"
plot_metrics(model_0, dataset, True)
```

Output saved by creating file at /content/images/model_0_pp.png.
 Output saved by creating file at /content/images/model_0_pp.txt.
 Basic Model for Set: dummy_mouse_enhancers_ensembl:
 Total Loss: 0.579737
 Accuracy: 66.82%
 F1 Score: 33.13

Final Model for Set: dummy_mouse_enhancers_ensembl:
 Total Loss: 0.999781
 Accuracy: 50.64%
 F1 Score: 0.76



Drosophila enhancers vs Random background regions from dm6

Dataset 1: drosophila_enhancers_stark

- The length of genomic intervals ranges from 236 to 3237, with average 2118.1238 and median 2142.0
- Totally 6914 sequences have been found, 5184 for training and 1730 for testing

In [24]:

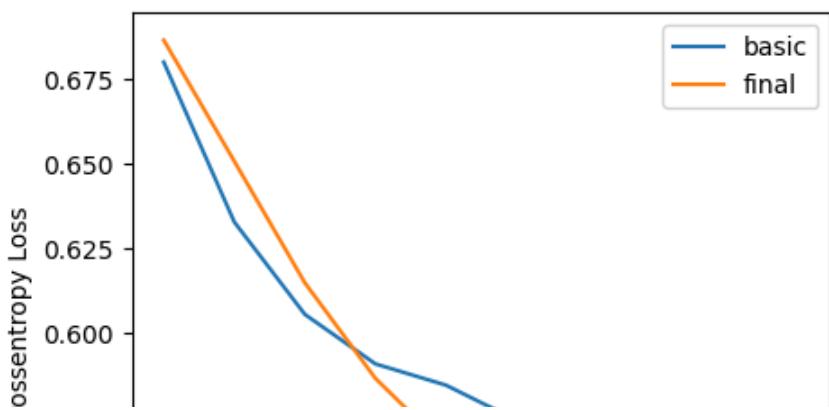
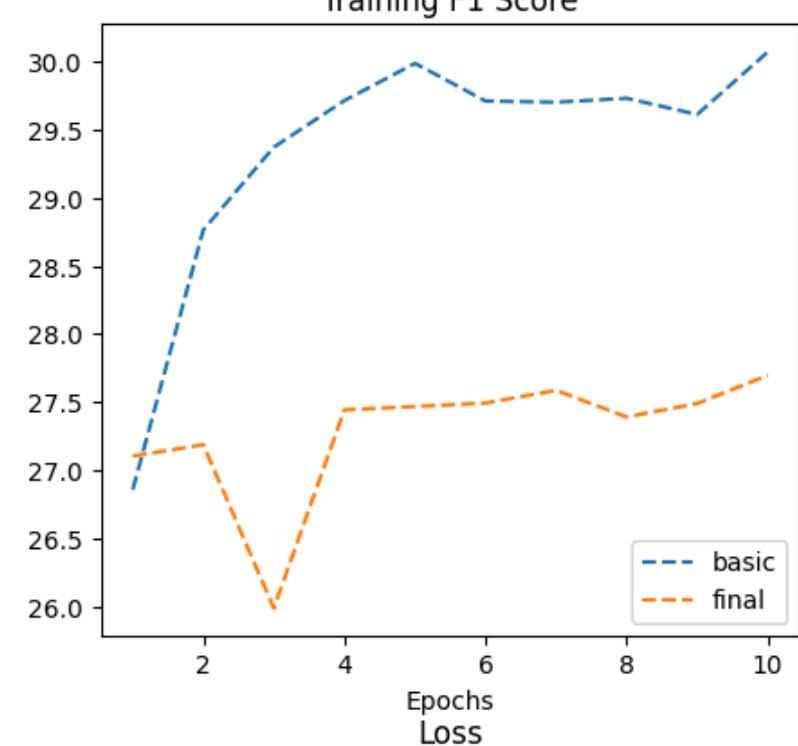
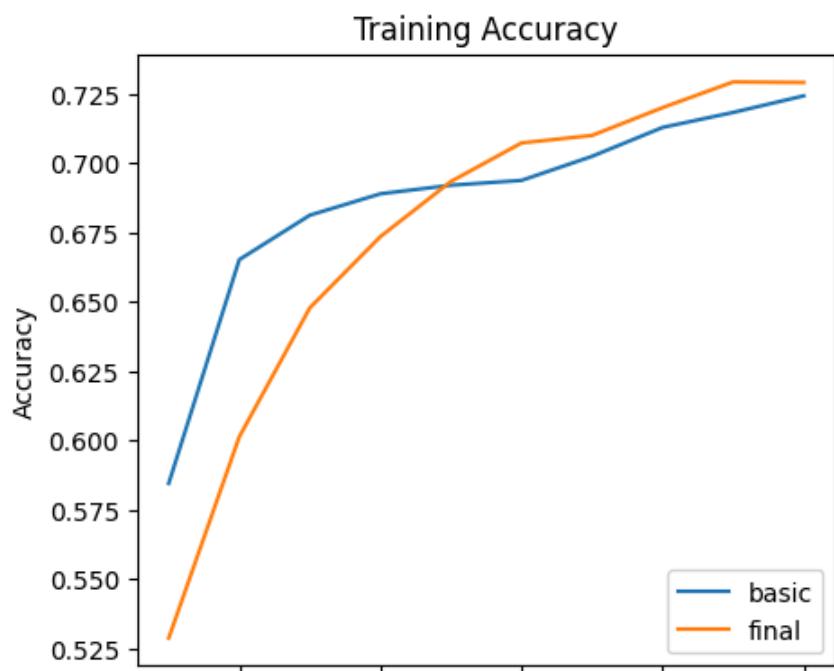
```
dataset = update_train_test_sets(1)
model_1 = train_and_evaluate_models(dataset, epochs)
```

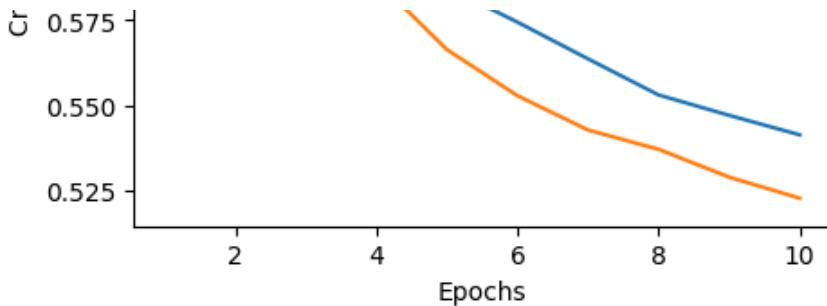
Found 5184 files belonging to 2 classes.
 Found 1730 files belonging to 2 classes.
 Training Models for: drosophila_enhancers_stark
 <Lambda name=lambda_2, built=False>

```
In [25]: %%capture_img --path "/content/images/model_1.png"
%%capture_text --path "/content/images/model_1.txt"
plot_metrics(model_1, dataset)

Output saved by creating file at /content/images/model_1.png.
Output saved by creating file at /content/images/model_1.txt.
Basic Model for Set: drosophila_enhancers_stark:
Total Loss: 1.369164
Accuracy: 49.94%
F1 Score: 40.93

Final Model for Set: drosophila_enhancers_stark:
Total Loss: 0.892226
Accuracy: 50.46%
F1 Score: 0.82
```





```
In [26]: # For presentation
```

```
%%capture_img --path "/content/images/model_1_pp.png"
plot_metrics(model_1, dataset, True)
```

Output saved by creating file at /content/images/model_1_pp.png.

Basic Model for Set: drosophila_enhancers_stark:

Total Loss: 1.369164

Accuracy: 49.94%

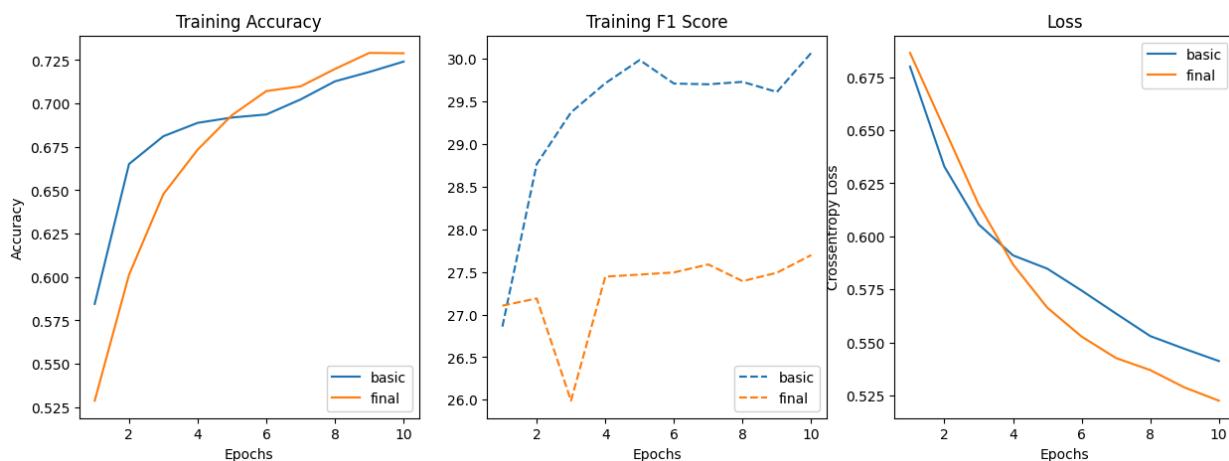
F1 Score: 40.93

Final Model for Set: drosophila_enhancers_stark:

Total Loss: 0.892226

Accuracy: 50.46%

F1 Score: 0.82



Human Enhancers

Dataset 2: *human_enhancers_cohn*

- All lengths of genomic intervals equals 500
- Totally 27791 sequences have been found, 20843 for training and 6948 for testing

```
In [27]: dataset = update_train_test_sets(2)
model_2 = train_and_evaluate_models(dataset, epochs)
```

Found 20843 files belonging to 2 classes.

Found 6948 files belonging to 2 classes.

Training Models for: *human_enhancers_cohn*

<Lambda name=lambda_4, built=False>

```
In [28]: %%capture_img --path "/content/images/model_2.png"
%%capture_text --path "/content/images/model_2.txt"
plot_metrics(model_2, dataset)
```

Output saved by creating file at /content/images/model_2.png.

Output saved by creating file at /content/images/model_2.txt.

Basic Model for Set: human_enhancers_cohn:

Total Loss: 0.792922

Accuracy: 52.6%

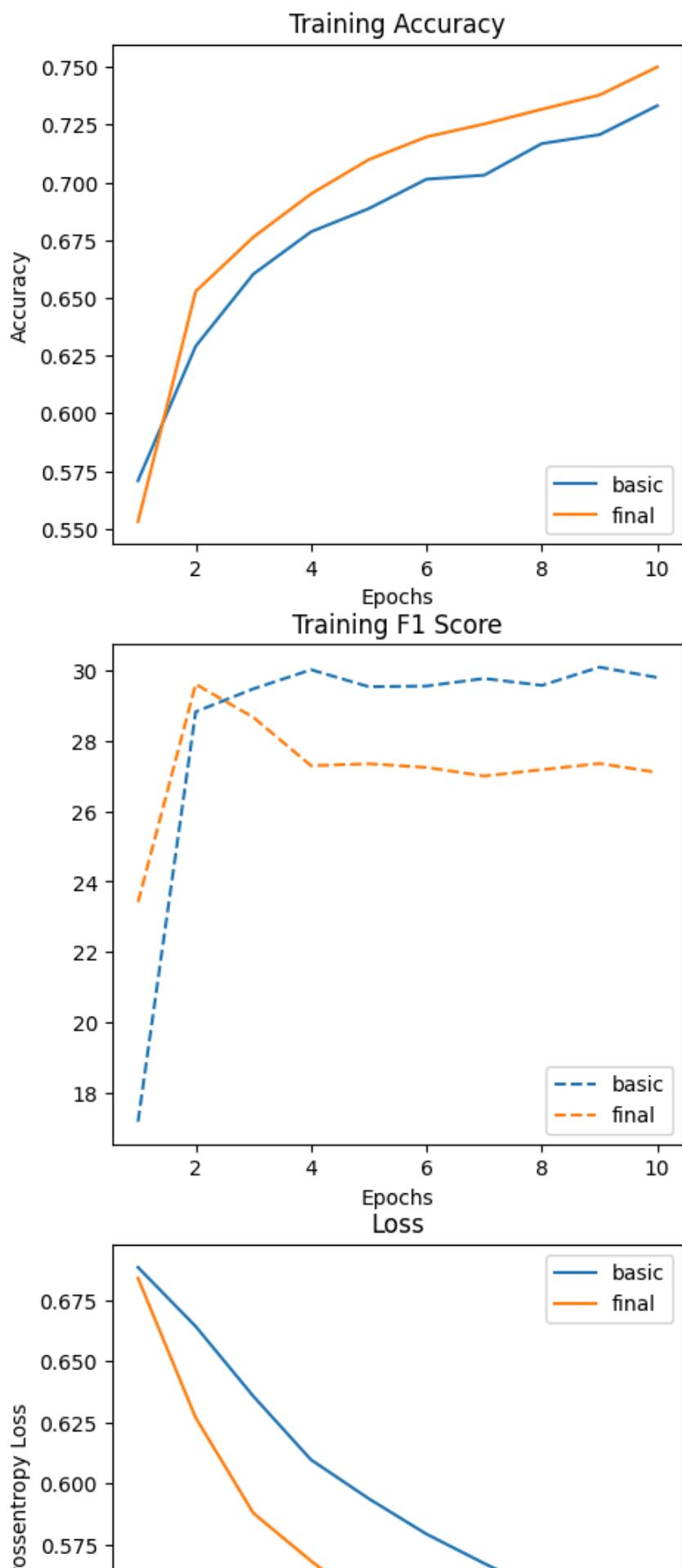
F1 Score: 40.45

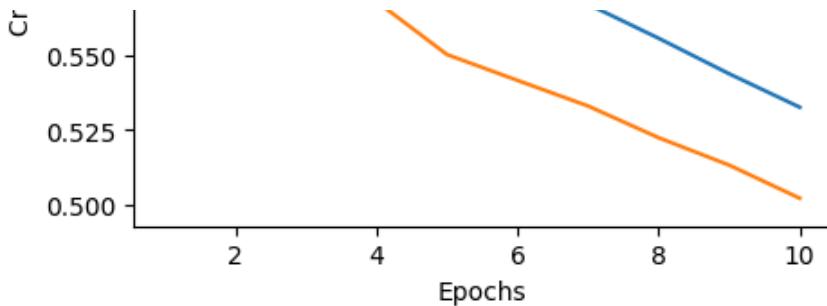
Final Model for Set: human_enhancers_cohn:

Total Loss: 0.626275

Accuracy: 65.09%

F1 Score: 14.97





In [29]: # For presentation

```
%capture_img --path "/content/images/model_2_pp.png"
plot_metrics(model_2, dataset, True)
```

Output saved by creating file at /content/images/model_2_pp.png.

Basic Model for Set: human_enhancers_cohn:

Total Loss: 0.792922

Accuracy: 52.6%

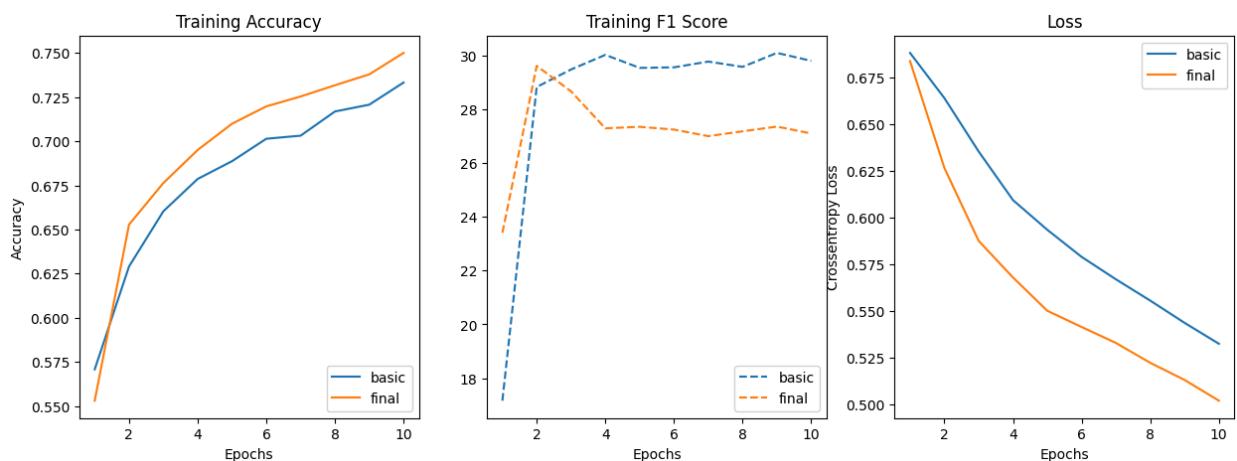
F1 Score: 40.45

Final Model for Set: human_enhancers_cohn:

Total Loss: 0.626275

Accuracy: 65.09%

F1 Score: 14.97



Human non-TATA promoters

Dataset 3: *human_nontata_promoters*

- All lengths of genomic intervals equals 251
- Totally 36131 sequences have been found, 27097 for training and 9034 for testing

In [30]: dataset = update_train_test_sets(3)
model_3 = train_and_evaluate_models(dataset, epochs)

Found 27097 files belonging to 2 classes.

Found 9034 files belonging to 2 classes.

Training Models for: human_nontata_promoters

<Lambda name=lambda_6, built=False>

In [31]: %capture_img --path "/content/images/model_3.png"
%capture_text --path "/content/images/model_3.txt"
plot_metrics(model_3, dataset)

Output saved by creating file at /content/images/model_3.png.

Output saved by creating file at /content/images/model_3.txt.

Basic Model for Set: human_nontata_promoters:

Total Loss: 0.581651

Accuracy: 71.56%

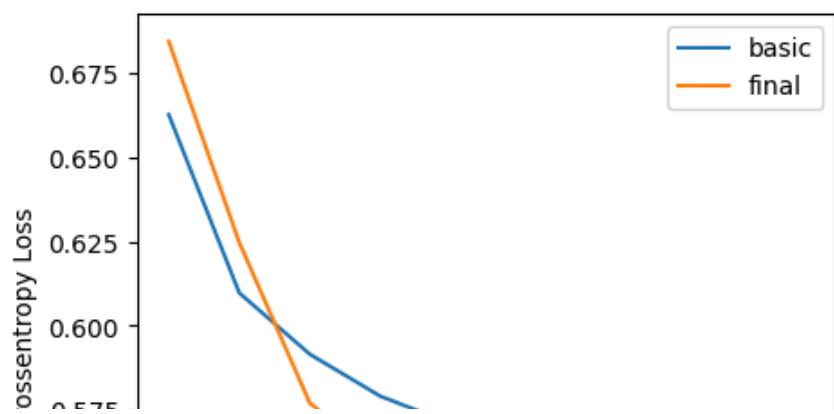
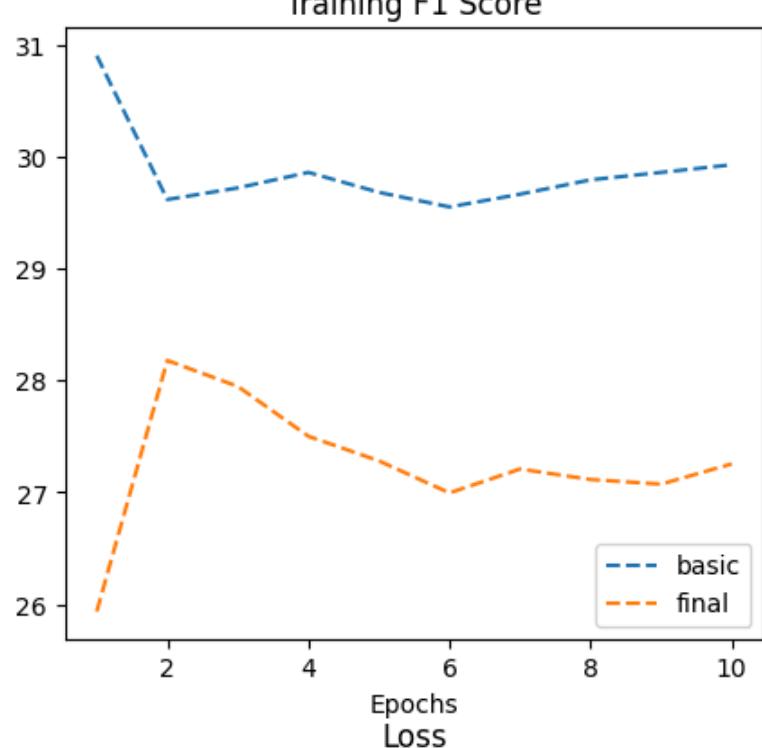
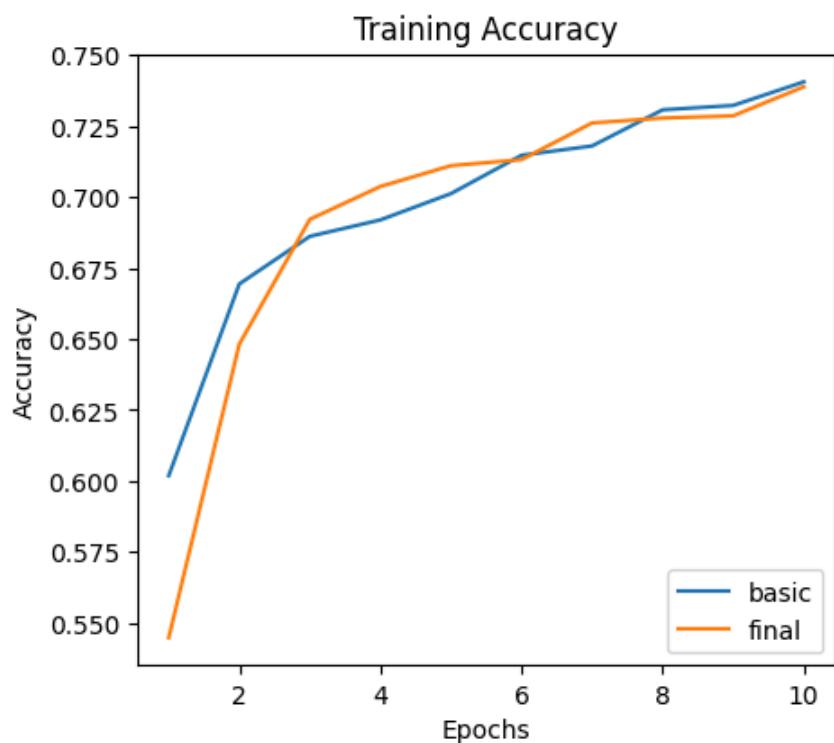
F1 Score: 23.3

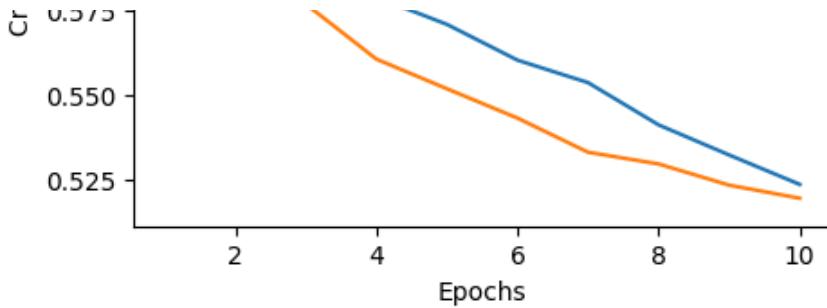
Final Model for Set: human_nontata_promoters:

Total Loss: 0.606759

Accuracy: 66.36%

F1 Score: 20.33





```
In [32]: # For presentation
```

```
%capture_img --path "/content/images/model_3_pp.png"
plot_metrics(model_3, dataset, True)
```

Output saved by creating file at /content/images/model_3_pp.png.

Basic Model for Set: human_nontata_promoters:

Total Loss: 0.581651

Accuracy: 71.56%

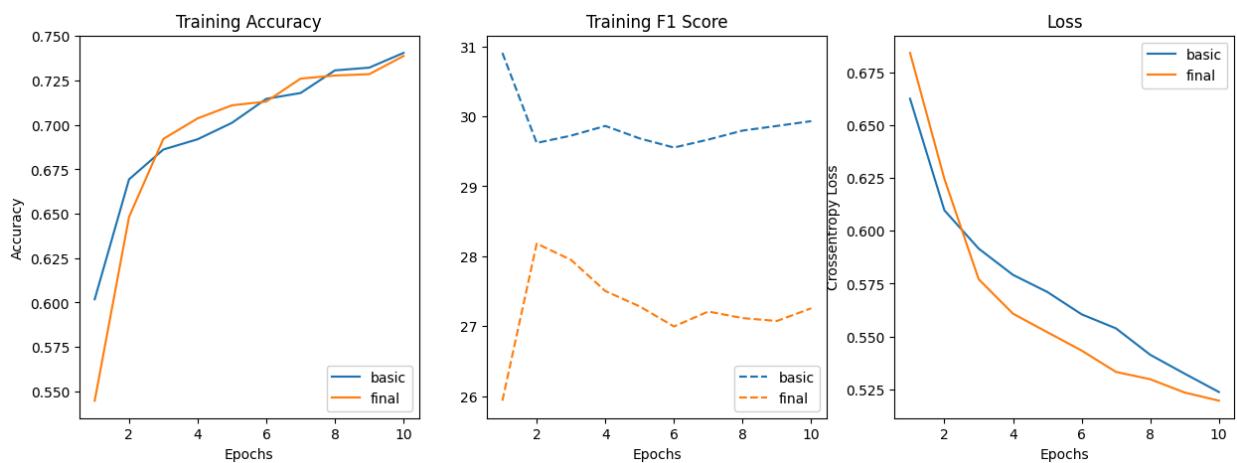
F1 Score: 23.3

Final Model for Set: human_nontata_promoters:

Total Loss: 0.606759

Accuracy: 66.36%

F1 Score: 20.33



Mitigation and Workarounds

The issues that could not be solve have been added to this text block so the commands do not execute.

Tensorflow-Addons Issue:

```
!pip install -q tensorflow-addons
```

After I was able to get the package install there was an issue with the Keras version and the removal of the engine object in the new version of Keras. Also a this would have only worked for the first tested model. So, a new *F₁ Score* metric was written instead.

Torchtext Issue:

There was problem with the compatibility of the torch version with torchtext. There also was a problem with the python and cuda versions with a older version of torch, but a reinstall was attempted.

```
import torch
print(torch.__version__)

Output:
2.6.0+cu124
!python --version

!sudo apt-get -qq install python3.11 \
    python3.11-distutils \
    python3-pip

!sudo update-alternatives --install /usr/local/bin/python3 \
    python3 \
    /usr/bin/python3.11 1

!sudo update-alternatives --config python3
!python --version
!pip uninstall torch torchtext -
!pip cache purge
# Example for a specific CUDA version (check PyTorch website for the exact
command)
!pip install torch==2.2.0 torchvision==0.17.0 torchtext==0.18.0
torchaudio==2.2.0 --index-url https://download.pytorch.org/whl/cu12
Output:
ERROR: Could not find a version that satisfies the requirement torchtext==0.18.0 (from versions: 0.5.0, 0.6.0,
0.15.0+cpu, 0.15.1+cpu, 0.15.2+cpu, 0.16.0+cpu, 0.16.1+cpu, 0.16.2+cpu, 0.17.0+cpu).
ERROR: No matching distribution found for torchtext==0.18.0
```

Genomic-benchmarks Issues:

These imports were could not be used do to *tensorflow-addons* package or the version update of keras.

```
from genomic_benchmarks.models.tf import (
    get_basic_cnn_model_v0,
    vectorize_layer,
)
```

These imports relied on the version of *torch* and *torchtext* packages from the above issue.

```
from genomic_benchmarks.dataset_getters.pytorch_datasets import (
    get_dataset,
)
from genomic_benchmarks.dataset_getters.utils import (
    build_vocab,
    check_seq_lengths,
    check_config,
    coll_factory,
    LetterTokenizer,
    VARIABLE_LENGTH_DATASETS,
)
from genomic_benchmarks.models.torch import CNN
```

Dataset Issue:

Testing moving the datasets from a prefetchdataset to a native *numpy* arrays.

```

import tensorflow_datasets as tfds
x1 = np.asarray(list(map(lambda x: x[0], tfds.as_numpy(train_ds))))
x1
Output:
-----
ValueError Traceback (most recent call last)
/tmp/ipython-input-80-1145905047.py in ()
1 import tensorflow_datasets as tfds
----> 2 x1 = np.asarray(list(map(lambda x: x[0], tfds.as_numpy(train_ds))))
3 x1
ValueError: setting an array element with a sequence. The requested array has an inhomogeneous shape after 1 dimensions. The detected shape was (16,) + inhomogeneous part.

```

Keras Issue (Solved):

The built-in *F1Score* in the metrics package in *keras* would return this error when the model was training. Output:

```

/usr/local/lib/python3.11/dist-packages/keras/src/metrics/f_score_metrics.py in _build(self, y_true_shape,
y_pred_shape)
122 def _build(self, y_true_shape, y_pred_shape):
123 if len(y_pred_shape) != 2 or len(y_true_shape) != 2:
--> 124 raise ValueError(
125 "FBetaScore expects 2D inputs with shape "
126 "(batch_size, output_dim). Received input "
ValueError: FBetaScore expects 2D inputs with shape (batch_size, output_dim). Received input shapes:
y_pred.shape=(None, 1) and y_true.shape=(None,).

```

Fix: A function was added to gather the *F1 Score* metric.

Jupyter Capture Issue:

The package was successful installed and the cell output capture is reporting a working status, but outputs are not appearing in the directory. *FIX: The capture magic methods did work for some image and text outputs.*

Mermaid-py Issue:

The package was cutting off text in the layer diagram. *FIX: I used the online editor at [Mermaid Live Editor](#)*

Mermaid Diagrams

```

In [47]: basic_cnn_graph = """
%{init:{'flowchart':{'nodeSpacing':1, 'rankSpacing':10}}}%
flowchart TD
    classDef withMargins fill-opacity:0.0,color:#FFFFFF,stroke-width:0px;
    i(Embedding)

    subgraph conv1[Convolution Layer 1]
        space1["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        c1@{ shape: st-rect, label: "Conv1D:<br>32 filters<br>kernel 8" }
        b1@{ shape: st-rect, label: "Batch Normalization" }
        p1@{ shape: st-rect, label: "Max Pooling" }

```

```

end
%% Define a class to make the padding subgraph invisible
classDef padding stroke:none,fill:none

subgraph conv2[Convolution Layer 2]
    space2["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargines;
    c2@{ shape: st-rect, label: "Conv1D:\n16 filters - kernel 8" }
    b2@{ shape: st-rect, label: "Batch Normalization" }
    p2@{ shape: st-rect, label: "Max Pooling" }
end

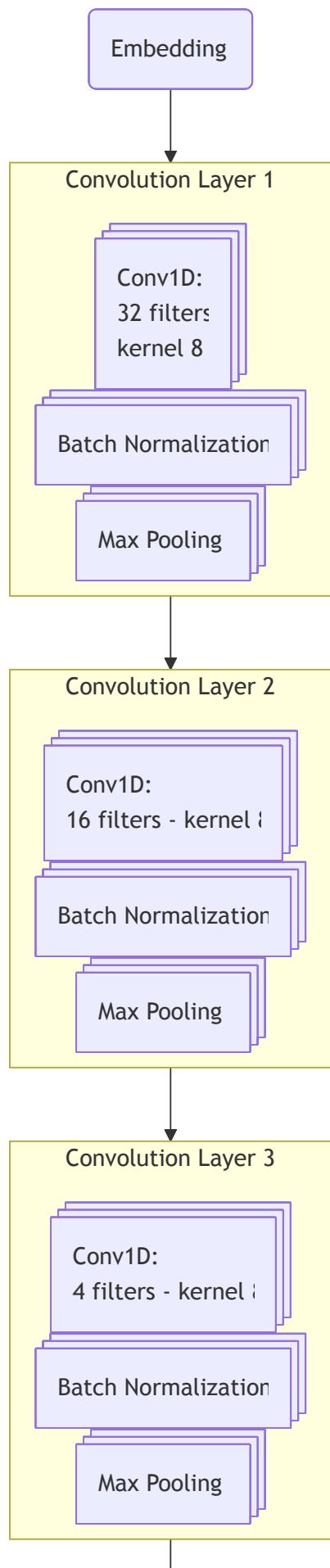
subgraph conv3[Convolution Layer 3]
    space3["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargines;
    c3@{ shape: st-rect, label: "Conv1D:\n4 filters - kernel 8" }
    b3@{ shape: st-rect, label: "Batch Normalization" }
    p3@{ shape: st-rect, label: "Max Pooling" }
end

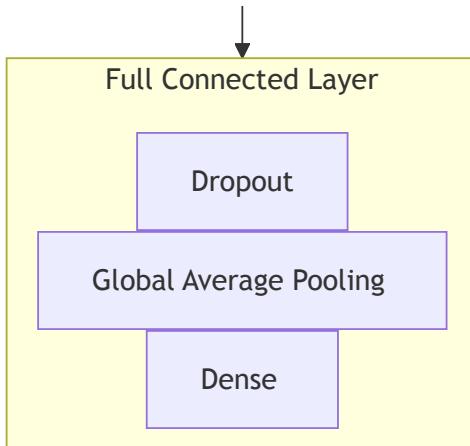
subgraph dense[Full Connected Layer]
    space4["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargines;
    do1[Dropout]
    p4[Global Average Pooling]
    d1[Dense]
end

i -----> conv1
conv1 -----> conv2
conv2 -----> conv3
conv3 -----> dense
....
```

Mermaid(basic_cnn_graph)

Out[47] :





```

In [49]: final_cnn_graph = """
%%{init:{'flowchart':{'nodeSpacing':1, 'rankSpacing':10}}}%

flowchart TD
    classDef withMargins fill-opacity:0.0,color:#FFFFFF,stroke-width:0px;
    i[Embedding]

    subgraph conv1[Convolution Layer 1]
        space1["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        c1@{ shape: st-rect, label: "Conv1D:\n16 filters – kernel 8" }
        b1@{ shape: st-rect, label: "Batch Normalization" }
        a1@{ shape: st-rect, label: "Activation ReLU" }
        p1@{ shape: st-rect, label: "Max Pooling" }
    end
    %% Define a class to make the padding subgraph invisible
    classDef padding stroke:none,fill:none

    subgraph conv2[Convolution Layer 2]
        space2["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        c2@{ shape: st-rect, label: "Conv1D:\n8 filters – kernel 8" }
        b2@{ shape: st-rect, label: "Batch Normalization" }
        p2@{ shape: st-rect, label: "Max Pooling" }
    end

    subgraph conv3[Convolution Layer 3]
        space3["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        c3@{ shape: st-rect, label: "Conv1D:\n4 filters – kernel 8" }
        b3@{ shape: st-rect, label: "Batch Normalization" }
        p3@{ shape: st-rect, label: "Max Pooling" }
    end

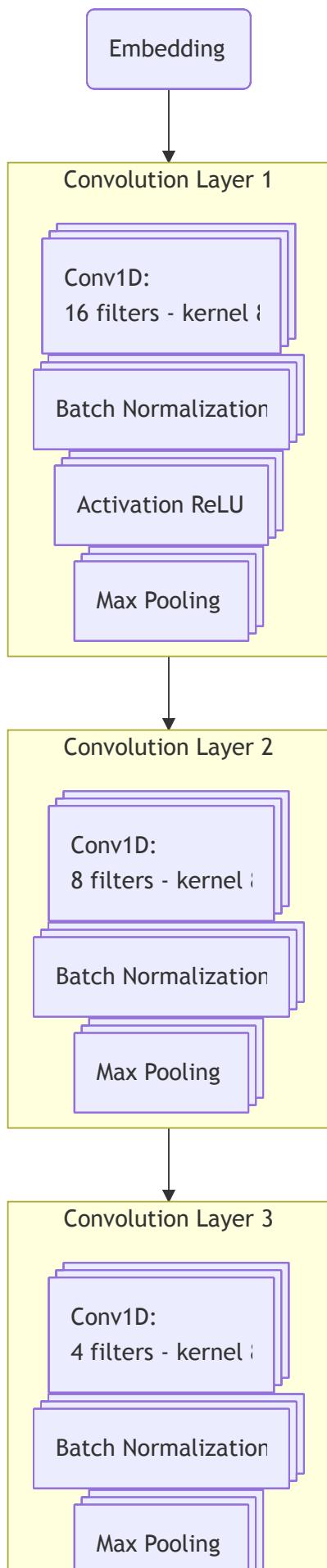
    subgraph dense[Full Connected Layer]
        space4["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        f[Flatten]
        d1[Dense]
        d2[Dense]
    end

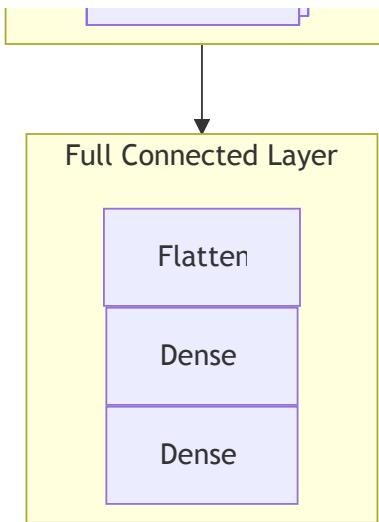
    i -----> conv1
    conv1 -----> conv2
    conv2 -----> conv3
    conv3 -----> dense
    .....

```

Mermaid(final_cnn_graph)

Out[49] :





```

In [73]: %%capture_code --path "/content/code/mermaid_example.py"
final_cnn_graph_update = """
%%{init:{'flowchart':{'nodeSpacing':1, 'rankSpacing':10}}}{}

flowchart TD
    classDef withMargins fill-opacity:0.0,color:#FFFFFF,stroke-width:0px;
    i(Embedding)

    subgraph conv1[Convolution Layer 1]
        space1["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        c1@{ shape: st-rect, label: "Conv1D:\n16 filters - kernel 8" }
        b1@{ shape: st-rect, label: "Batch Normalization" }
        a1@{ shape: st-rect, label: "Activation ReLU" }
        p1@{ shape: st-rect, label: "Max Pooling" }
    end
    % Define a class to make the padding subgraph invisible
    classDef padding stroke:none,fill:none

    subgraph conv2[Convolution Layer 2]
        space2["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        c2@{ shape: st-rect, label: "Conv1D:\n8 filters - kernel 8" }
        b2@{ shape: st-rect, label: "Batch Normalization" }
        p2@{ shape: st-rect, label: "Max Pooling" }
    end

    subgraph conv3[Convolution Layer 3]
        space3["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        c3@{ shape: st-rect, label: "Conv1D:\n4 filters - kernel 8" }
        b3@{ shape: st-rect, label: "Batch Normalization" }
        p3@{ shape: st-rect, label: "Max Pooling" }
    end

    subgraph dense[Full Connected Layer]
        space4["<p style='width:100px;height:0px;margin:0'>Space</p>"]:::withMargins;
        p4[Global Average Pooling]
        f[Flatten]
        d1[Dense]
        d2[Dense]
        f --- d1
        d1 --- d2
    end

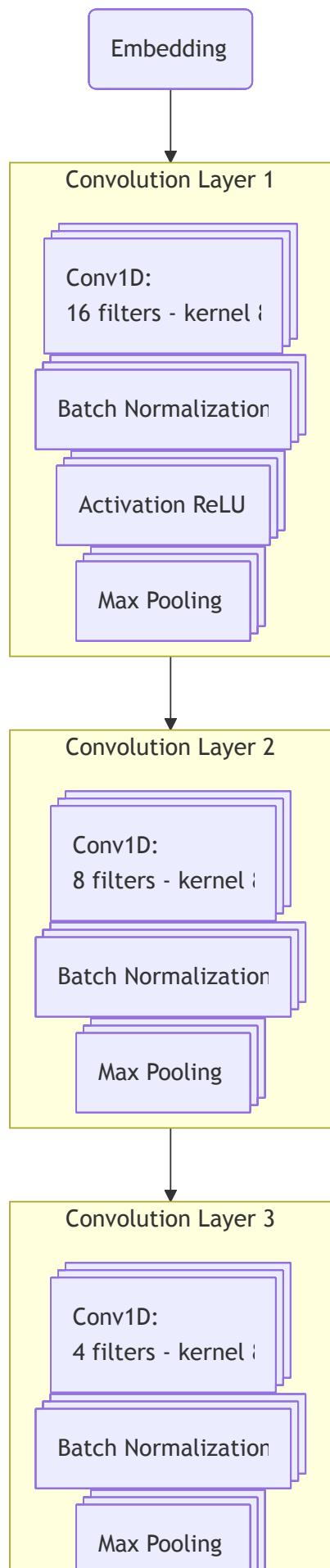
    i -----> conv1
    conv1 -----> conv2
    conv2 -----> conv3

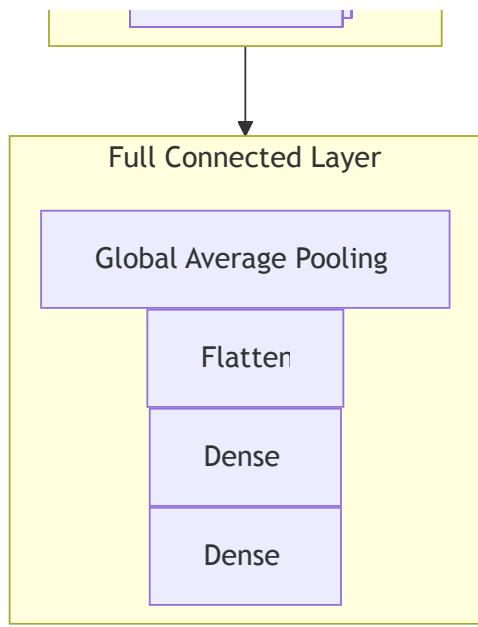
```

```
conv3 -----> dense  
....  
Mermaid(final_cnn_graph_update)
```

Output saved by creating file at /content/code/mermaid_example.py.

Out[73]:





End of Project