```
import tensorflow_decision_forests as tfdf
import os
import numpy as np
import pandas as pd
import tensorflow as tf
import math
from IPython.core.magic import register_line_magic
from IPython.display import Javascript
from IPython.display import display as ipy_display
import rasterio
import pandas as pd
from sklearn.model_selection import train_test_split
# Some of the model training logs can cover the full
# screen if not compressed to a smaller viewport.
# This magic allows setting a max height for a cell.
@register_line_magic
def set_cell_height(size):
  ipy_display(
      Javascript("google.colab.output.setIframeHeight(0, true, {maxHeight: " +
                str(size) + "})"))
# Check the version of TensorFlow Decision Forests
print("Found TensorFlow Decision Forests v" + tfdf.__version__)
     Found TensorFlow Decision Forests v1.8.1
grupo_path = '/content/drive/MyDrive/data/output_grupo_raster.tif'
binary_path = '/content/drive/MyDrive/data/binary_deforestation_raster.tif'
# Load the raster data
with rasterio.open(grupo_path) as src:
    feature_raster = src.read(1)
with rasterio.open(binary_path) as src:
    label_raster = src.read(1)
# Flatten the raster arrays and create a DataFrame
pixels = {'feature': feature_raster.flatten(),
          'label': label_raster.flatten()}
df = pd.DataFrame(pixels)
# Remove nodata pixels (assuming nodata is -1 for label)
df = df[df['label'] != -1]
# Split the data into training and testing sets
train_df, test_df = train_test_split(df, test_size=0.3)

★ Generate

                 Using ...
                            can we remove pixels and df from memory or is it still needed?
                                                                                                                                            Q
                                                                                                                                                  Close
 del df
del pixels
# Convert DataFrame to TensorFlow dataset
train_ds = tfdf.keras.pd_dataframe_to_tf_dataset(train_df, label='label')
test_ds = tfdf.keras.pd_dataframe_to_tf_dataset(test_df, label='label')
```

```
# Train a Random Forest model
model = tfdf.keras.RandomForestModel()
model.fit(train_ds)
    Use /tmp/tmp8qr8ocr2 as temporary training directory
    Reading training dataset...
    Training dataset read in 0:01:27.313504. Found 59064655 examples.
    Training model...
    Model trained in 0:24:07.327010
    Compiling model...
    Model compiled.
    <keras.src.callbacks.History at 0x78da5c7146d0>
# Evaluate the model
evaluation = model.evaluate(test_ds)
print(evaluation)
    0.0
model.compile(metrics=["accuracy"])
evaluation = model.evaluate(test_ds, return_dict=True)
print()
for name, value in evaluation.items():
 print(f"{name}: {value:.4f}")
    loss: 0.0000
    accuracy: 0.7076
tfdf.model_plotter.plot_model_in_colab(model, tree_idx=0, max_depth=3)
                                                                                (6222064)
                                               feature >= 5.50000
                                                           (6948042)
                          feature >= 4.50000
                                      (30168411)
      feature >= 0.00000
                 (59064655)
                                      (28896244)
                                                                   feature >= 3.50000
                                               feature >= 2.50000
                                                           (23220369)
```

(19365093)

%set\_cell\_height 300
model.summary()

```
trees: 263, Out-o+-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 265, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 268, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 270, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 272, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 274, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 277, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 279, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 282, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 284, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 286, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 289, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 291, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 294, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 297, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
trees: 300, Out-of-bag evaluation: accuracy:0.707625 logloss:10.5383
```

Accuracy: The OOB accuracy of 0.707625 suggests that about 70.76% of the OOB samples were correctly classified by the model. Logloss: The logarithmic loss (logloss) of 10.5383 is a measure of error and is more sensitive to classifiers that are confident about an incorrect classification. OOB evaluation is a method of measuring the prediction error of random forest models. For each tree, it uses only the data that was not included in the bootstrap sample (the "out-of-bag" data) to evaluate the model's performance. This provides a good estimate of how well the model might perform on unseen data.