

Data

- The data used comes from the provided dataset.
- Only the image data from the dataset is being used. The location and year data has been dropped as it shows low correlation with the labels, and, it did not increase the performance of the model.
- Data augmentation is being performed through horizontal and vertical flips of the original images
- The data is also being cached and prefetched so as to increase the tensorflow models speed at processing it

Features Correlation Matrix

	label	latitude	longitude	year
label	1.000000	-0.087783	0.001103	0.066380
latitude	-0.087783	1.000000	-0.439139	-0.076184
longitude	0.001103	-0.439139	1.000000	0.192967
year	0.066380	-0.076184	0.192967	1.000000

Procedure Used

- OT Create a Neural Network using Transfer Learning selecting a pre-trained base model, adding Dense and Dropout layers on top to classify correctly avoiding overfitting
- Train the top layers of the model tracking the loss output of the validation data for early stopping and reducing the learning rate of the model through exponential decay.
- Train the entire model with a smaller learning rate to avoid overfitting, implementing the same features as before.
- Repeat the steps above with an 80-20 split of train and validation data tuning the hyperparameters of the model. Once a model is chosen, train it with the entire dataset in order to make future predictions.

Model structure and results

As base model, we will be using the Inception V3 Network, from tensorflow hub (https://tfhub.dev/google/inaturalist/inception_v3/feature_vector/5), pre-trained on the iNaturalist Dataset.

It yields a 0.804 f1_macro score on the validation data with a 80-20 split

Output Shape	Param # ======= 0
[(None, 332, 332, 3)]	
(None, 2048)	21802784
(None, 2048)	0
(None, 100)	204900
(None, 100)	0
(None, 100)	10100
(None, 3)	303
	[(None, 332, 332, 3)] (None, 2048) (None, 2048) (None, 100) (None, 100) (None, 100)

Total params: 22,018,087 Trainable params: 215,303

Non-trainable params: 21,802,784