Bike Riding Style Classification

DATA 5100, Image Classification Project

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Agenda

Introduction

- Purpose
- Analytical Approach
- Libraries
- Timeline of Analysis

Preparation

- Data Collection
- Data Preparation

EDA

- Data Block Overview
- Example Images

Modeling

- Epochs Parameters
- Confusion Matrix

Evaluation

- Deployment
- Final Evaluation

Introduction

- Purpose
- Domain Problem
- Analytic Approach

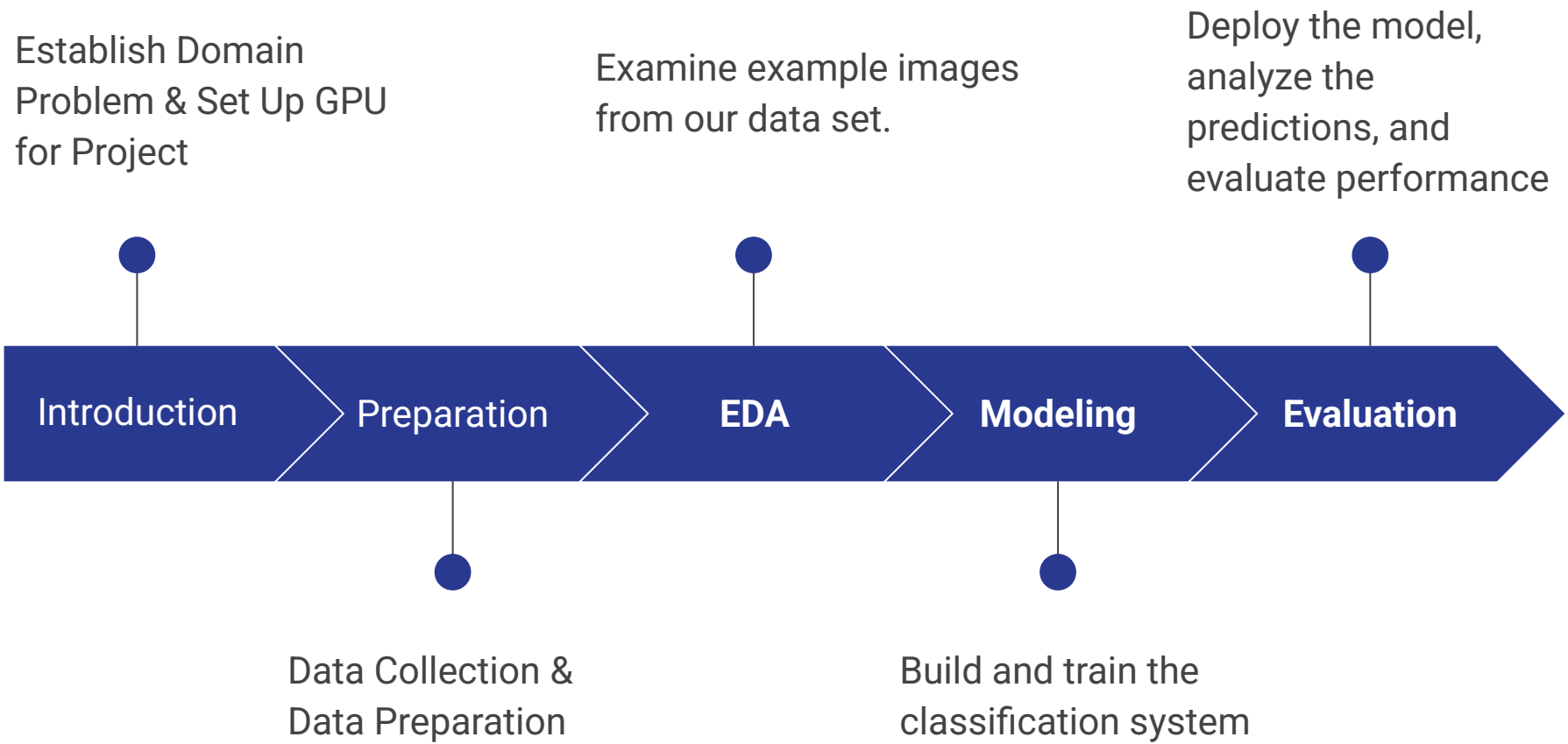
Introduction

Domain Problem

We want to develop a model to classify a bike based off of what it can see an image, into one of the following riding styles: **Cross Country**, **Downhill**, **Enduro**, **Trail** or **Dirt Jumper**.

Libraries

- **Kagglehub:** For downloading datasets directly from Kaggle into Google Colab.
- **Json:** For parsing and processing structured data from JSON files.
- **Pandas**: For cleaning, merging, and manipulating tabular data.
- **Matplotlib.pyplot**: For creating static visualizations and plotting model results.
- **Seaborn**: For statistical data visualization and probability plots.
- **Os**: For file and directory operations.
- **Requests**: For retrieving and downloading image data from URLs.
- **Tqdm**: For displaying progress bars during data download and preprocessing loops.
- **FastAI**: For training and testing the model - mainly the vision_learner and Classification Interpretation features


Establish Domain Problem & Set Up GPU for Project
Examine example images from our data set.
Deploy the model, analyze the predictions, and evaluate performance
Introduction
Preparation
EDA
Modeling
Evaluation
Data Collection & Data Preparation
Build and train the classification system

Preparation

- Data Collection
- Data Preparation

Data Collection

Bike Ads (images, prices, specifications)

- 10,052 bicycle advertisements from June 2020.
- 4 main data folders
  - `combined_price-only.csv`
  - `data_bike_exchange.json`
  - `data_ebay.json`
  - `images`

Data Preparation: Riding Style Classification

- Original Riding Styles

```
bikex_df['Riding Style'].value_counts()
```

Riding Style	count
Trail	364
Cross Country	322
Recreational	123
All Mountain	26
Enduro	25
Downhill & Freeride	19
Dirt Jump	10

dtype: int64

- New Riding Styles

```
df_stage2['Riding Style New'] = df_stage2['Riding Style'].apply(classify_stage2)
df_stage2['Riding Style New'].value_counts()
```

Riding Style New	count
trail	390
cross_country	322
dirt_jumper	133
enduro	25
downhill	19

dtype: int64

Exploratory Data Analysis

- Data Block Overview
- Example Images

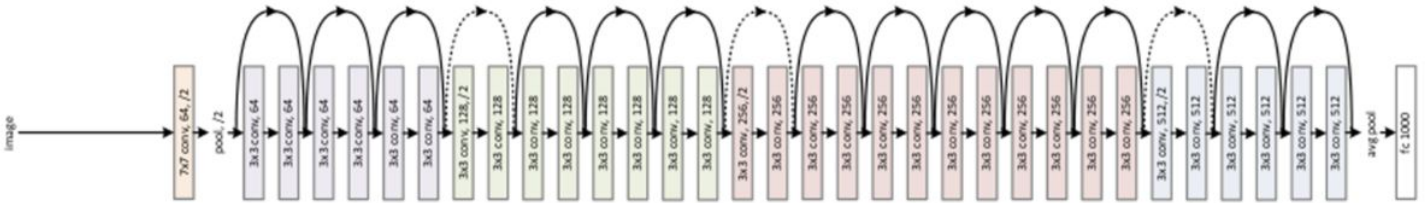
Example Images

```
# Check 12 labeled images
dls.show_batch(max_n=12, figsize=(10,8))
```

Modeling

- Epochs Parameters
- Confusion Matrix

ResNet-34 Modeling



Training & Testing

```
learn.fine_tune(epochs=1)
```

epoch	train_loss	valid_loss	accuracy	time
0	2.617749	1.485251	0.470588	00:06

epoch	train_loss	valid_loss	accuracy	time
0	1.755336	1.652832	0.470588	00:04

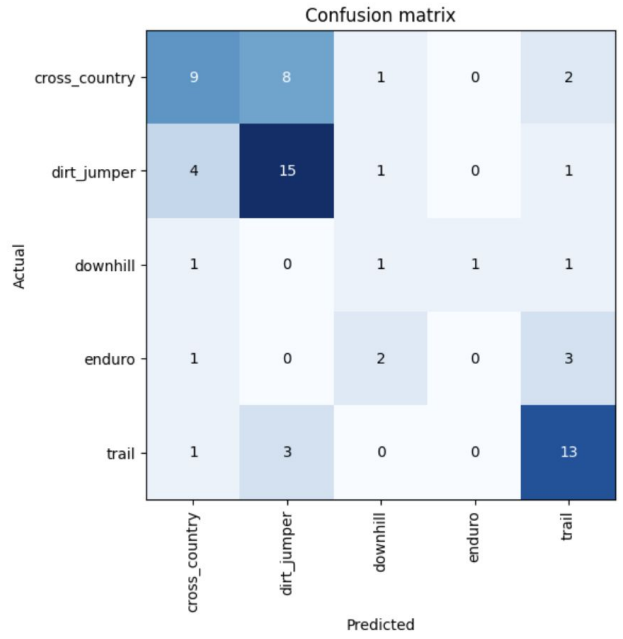
```
learn.fine_tune(3)
```

epoch	train_loss	valid_loss	accuracy	time
0	1.573802	1.350078	0.514706	00:04

epoch	train_loss	valid_loss	accuracy	time
0	1.527996	1.197022	0.602941	00:03
1	1.402186	1.321990	0.558824	00:04
2	1.314928	1.298977	0.558824	00:03

Confusion Matrix

```
[ ]
interp.plot_confusion_matrix(figsize=(6,6))
```



Evaluation

- Deployment
- Final Evaluation

Deployment, First Test

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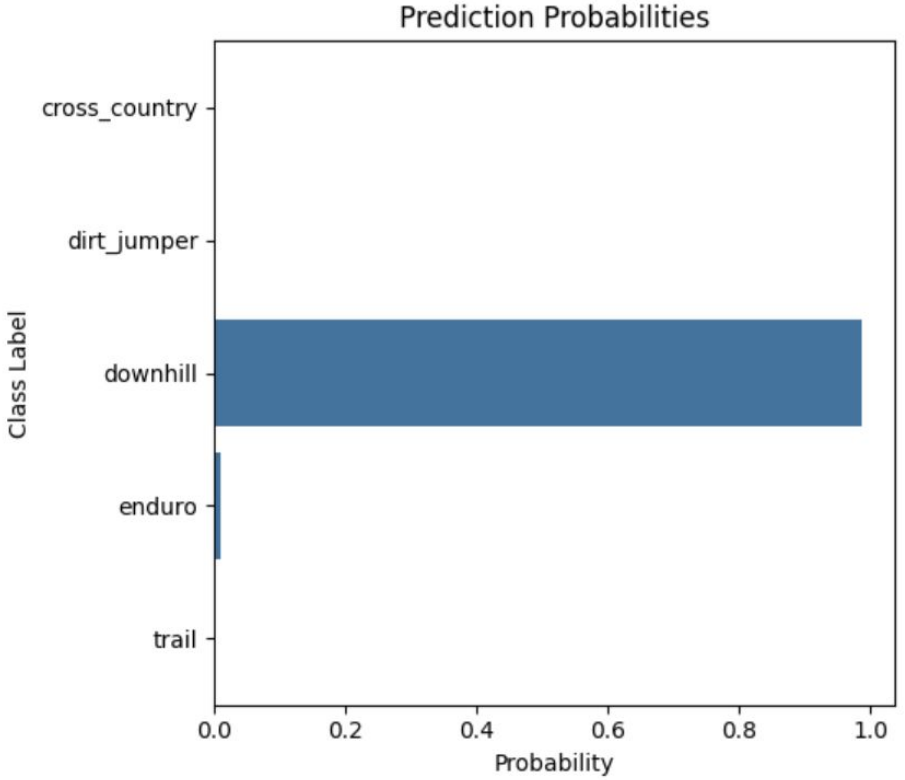
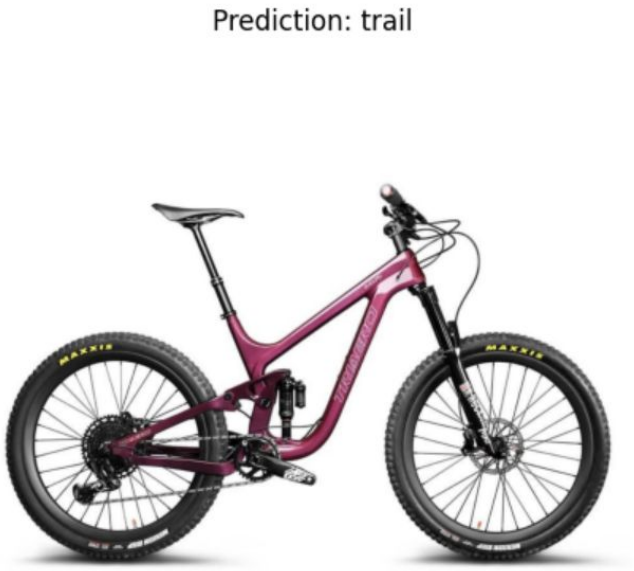
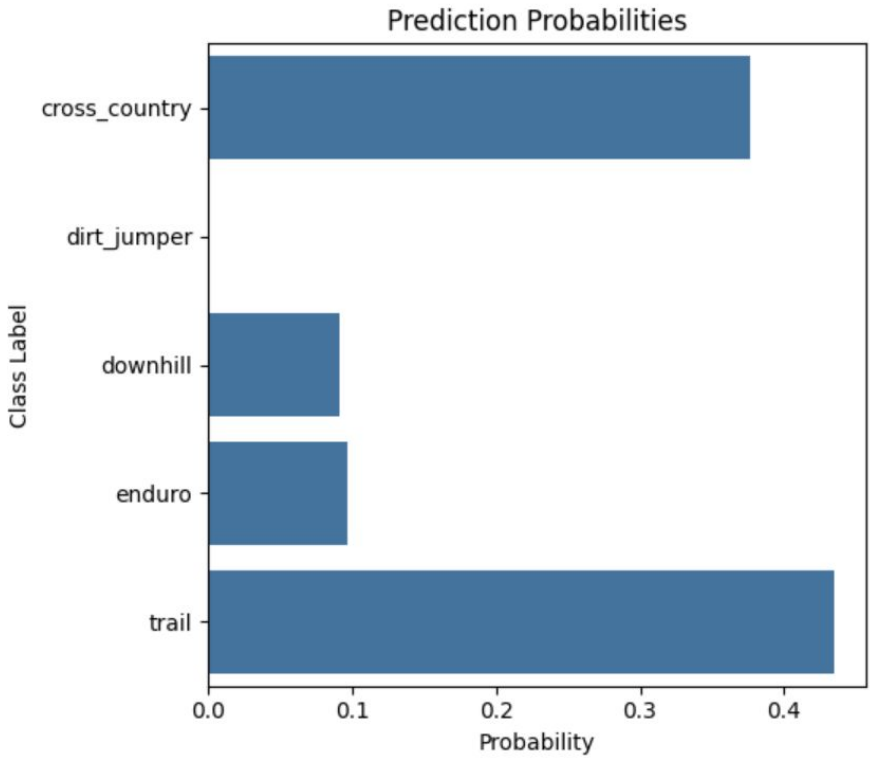


Deployment, Image 1

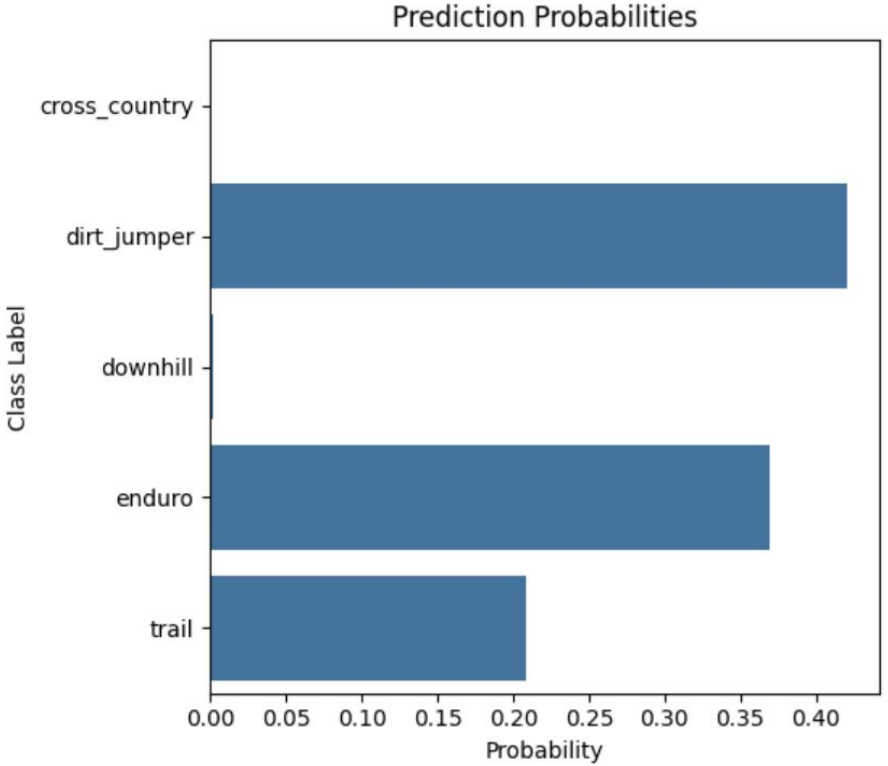
...





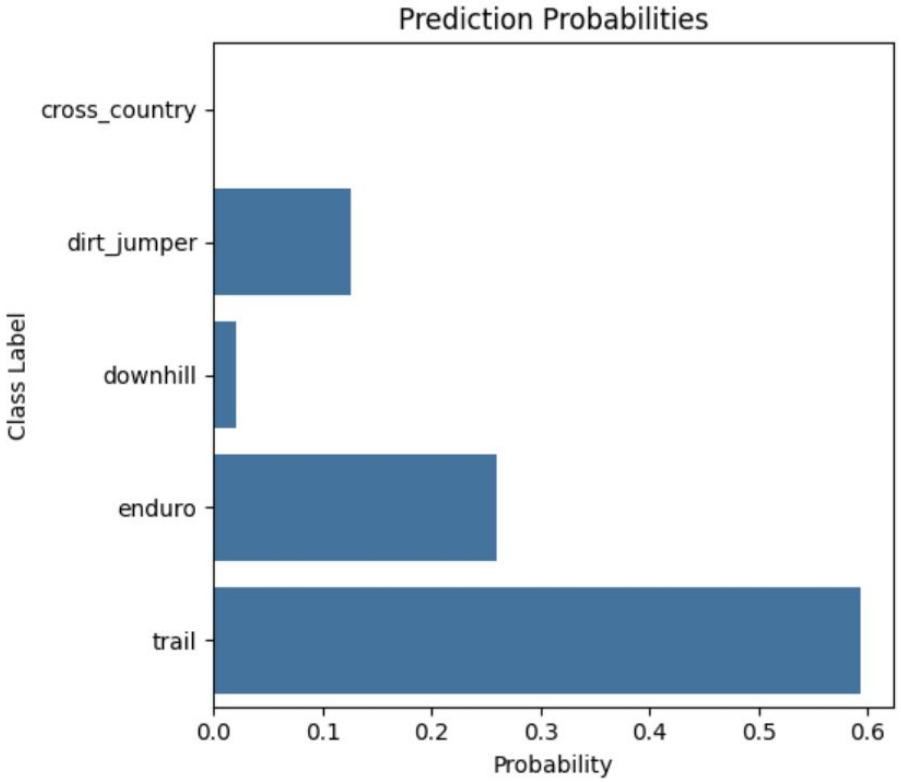
Deployment, Image 2

...



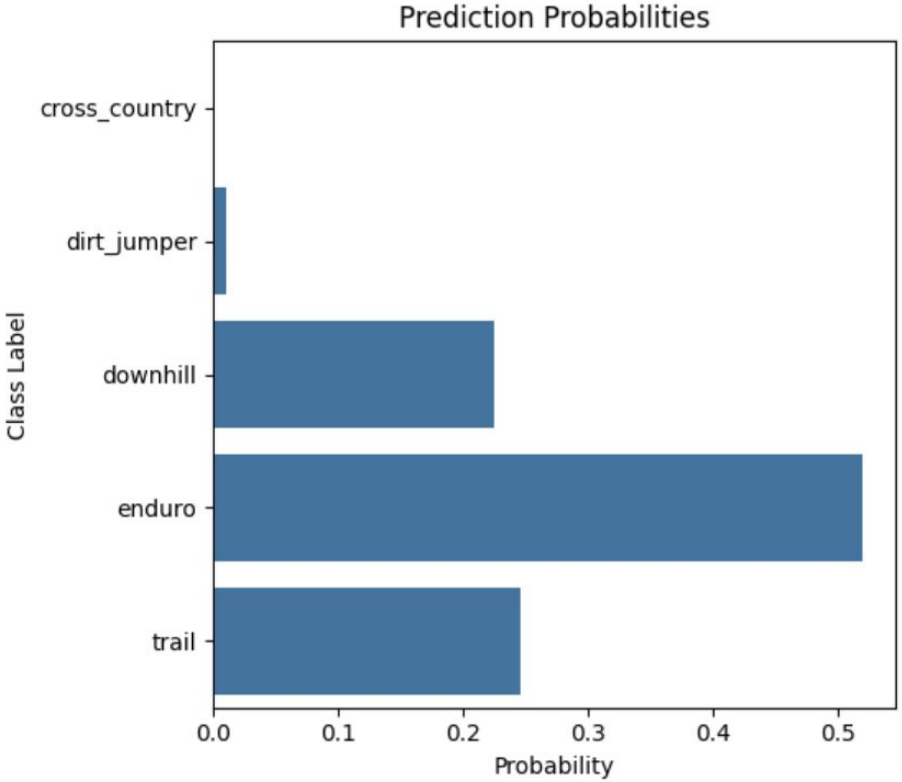
Deployment, Image 3

…



Deployment, Image 4

...



Final Evaluation

Domain Problem:

We want to develop a model to classify a bike based off of what it can see an image, into one of the following riding styles: **Cross Country**, **Downhill**, **Enduro**, **Trail** or **Dirt Jumper**.

Expansions for Improved Accuracy:

I believe the biggest things that could help make the model more accurate would be **limiting the image amount for each category.**