

Coin Analysis And Classification

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# Coin Detection Pipeline

#### Watershed segmentation

- Convert image to grayscale and apply Gaussian blur
- Apply Otsu's thresholding to separate coins from background
- Use morphological opening to remove small noise
- Dilate the image to get 'sure background' regions
- Apply distance transform to find 'sure foreground' (coin centers)
- Label regions and create markers
- Run Watershed algorithm to segment overlapping coins
- Extract and crop individual coins from the original image
- Save coin crops for further classification or analysis



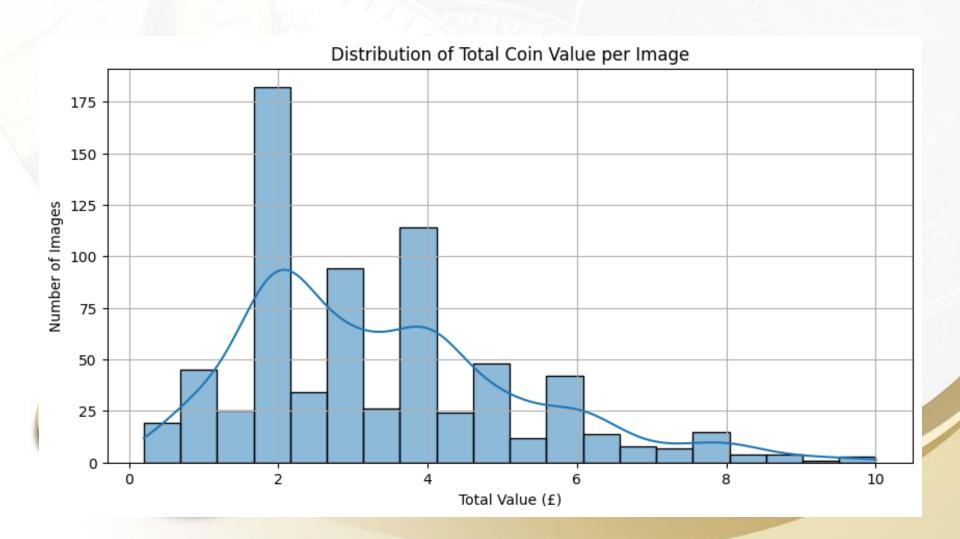
# Estimating UK Coin Value Per Image

#### **Heuristic Approach**

- Applied a rule-based classifier to estimate coin type
- Converted each coin to HSV color space
- Used average hue and coin size (diameter) to classify:
- Silver + small → likely 5p, 10p
- Reddish + small → likely 1p, 2p
- Mapped each coin to its monetary value
- Summed values by original image to estimate total value
- Plotted a histogram showing value distribution across images
- Quick, interpretable results without training a model



## Visualization



## CNN Model for Heads/Tails Classification

#### **Transfer Learning**

- Used MobileNetV2 as a pretrained base model (ImageNet)
- Removed original output layers and added a custom binary classifier
- Architecture:
- Global Average Pooling → Dense Layer → Dropout → Sigmoid output
- Base model frozen (trainable=False) to avoid overfitting
- Trained on a small labeled dataset of cropped coin images
- Efficient, accurate setup for quick training on limited data
- Can optionally be fine-tuned later for higher performance

### Manual Labeling Tool (For Training Data)

- Built a **simple interactive loop** to label cropped coin images
- Labeled up to 150 coins manually as "heads" or "tails"
- Used keyboard input to assign labels (heads, tails, skip, back, quit)
- Saved all results in a clean CSV file for training



```
abel as (heads / tails / skip / back / quit): heads

    Coin 127/1107 - coins140 coin4.png

👉 Label as (heads / tails / skip / back / quit): heads

    Coin 128/1107 - coins141 coin1.png

   Label as (heads / tails / skip / back / quit): tails

② Coin 129/1107 - coins141_coin2.png

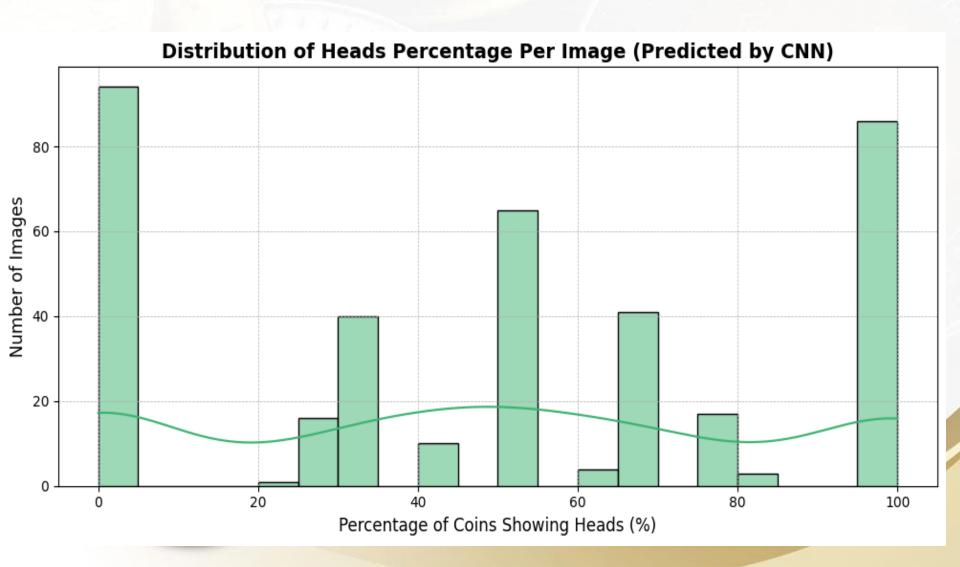
   Label as (heads / tails / skip / back / quit): skip
```

#### **Model Inference**

- •Ran the trained model on unlabeled coin crops
- Model predicted heads or tails for each coin
- •Grouped predictions by the original image each coin came from
- Calculated % of coins showing heads per image
- Plotted a histogram to visualize distribution
- Some images mostly heads, some mostly tails, some mixed



## Visualization



# Challenges

Unlabeled Dataset:

Had to manually label coin images for training and evaluation

Overlapping Coins:

Difficult to segment touching coins accurately

Unclear Coin Faces:

Some images were low-quality or blurry, making heads/tails labeling difficult

Lack of Pretrained Models:

Very few existing models for UK coin classification or orientation

Small Labeled Sample:

Only ~150 labeled coins available for training the CNN

Class Imbalance:

Uneven distribution of heads vs tails in labeled data

Visual Similarity Between Classes:

Heads and tails often look similar under poor lighting or angle

# Thank you!:)

