Coin Year Classification

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ELEN 520: Spring 2019

Problem Statement

- We will attempt to identify coin years from images of U.S. pennies
- H₀: Image pixel values are not correlated with the year of the coin
- H₁: Image pixel values are correlated with the year of the coin



Background

- Worth of coin is determined by year and condition
- Everyone has loose change.
 Could be valuable
- Ongoing debate about phasing out pennies

Ultra-Rare Penny Sells for \$204,000 at Auction

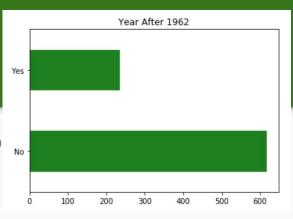
FORTUNE

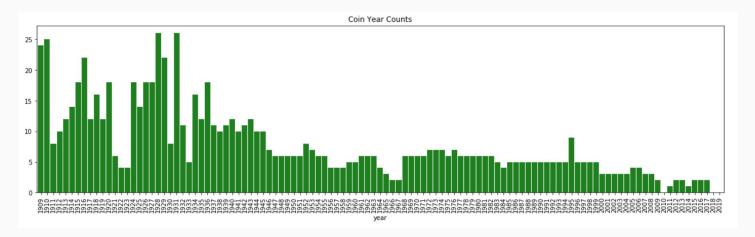
Chris Morris



Target Variable

- We chose to predict whether a coin is made befo or after 1963.
- The alternative is a target with 111 classes





Data

- Training data: We found 845 labeled penny images online
- Training data was scraped with the Selenium python package
 - o Image
 - Year
- Test data was collected with camera connected to PYNQ board



1959 Lincoln Memorial Cent Value



1959-D Lincoln Memorial Cent Value



1960 Lincoln Memorial Cent Value



1960-D Lincoln Memorial Cent Value



1961 Lincoln Memorial Cent Value



1961-D Lincoln Memorial Cent Value



1962 Lincoln Memorial Cent Value



1962-D Lincoln Memorial Cent Value



1963 Lincoln Memorial Cent Value



1963-D Lincoln Memorial Cent Value



1964 Lincoln Memorial Cent Value



1964-D Lincoln Memorial Cent Value

Scraping Images with Python

▶ <head>...</head>

▼<body>

1963 Lincoln Penny

CoinTrackers.com has estimated the 1963 Lincoln Penny value at an average of 1 cent, certified mint state (MS+) could be worth \$15. (see details)...

THE RIGHT CAR 3.0 PROVIDENT Img 260×210 Padding 5px 60px 5px 5px

Type: <u>Lincoln Penny</u> Year: 1963

Mint Mark: No mint mark Face Value: 0.01 USD

Total Produced: 754,110,000 [?]

Silver Content: 0%

Numismatic Value: 1 cent to \$15.00 Value: As a rough estimate of this coins value you can assume this coin in average condition will be valued at somewhere around 1 cent, while one in certified mint state (MS+) condition could bring as much as \$15 at auction. This price does not reference any standard coin grading scale. So when we say average, we mean in a

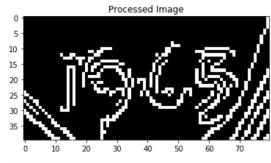


```
▼<div class="container">
▶ <div class="header">...</div>
▶ <div class="navigation">...</div>
▼<div class="main">
 ▼<div class="content">
   ▶ <div id="breadcrumbs">...</div>
    <h1>1963 Lincoln Penny</h1>
   ▶ ...
   ▶ ...
   ▼<div class="right">
      <img src="/img/coins/1963-lincoln-penny.png"</pre>
     alt="1963 Lincoln Penny" style="padding-right:
     60px;"> == $0
      <br>
       # Extract Image URL
       image xpath = '//*/div[1]/div[3]/div[1]/div[2]/img'
       image = driver.find element(By.XPATH,image xpath)
       image url = image.get attribute("src")
       # Save Image
       destination = 'ct/' + name + '.png'
      try:
           urllib.request.urlretrieve(image url, destination)
       except:
            pass
```

Image Preprocessing

- All operations performed by OpenCV
 - Apply gaussian blur
 - Canny edge detection
 - Slice matrix to include year location
- OpenCV is an open source package that allows complex image processing operations with only a few lines of code
- Size reduced from 50 x 150 to 40 x 80

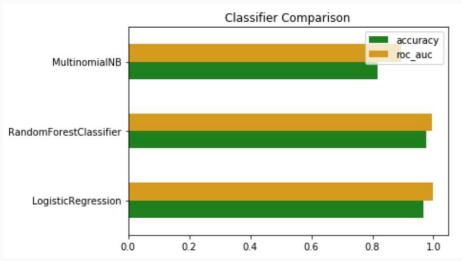




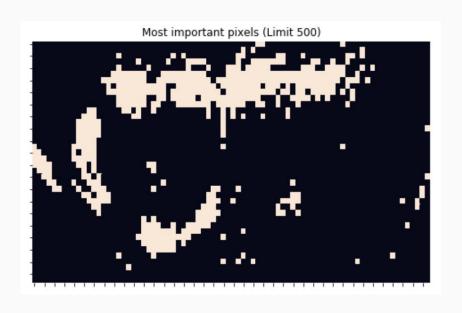
Classifier Comparison - Training Data

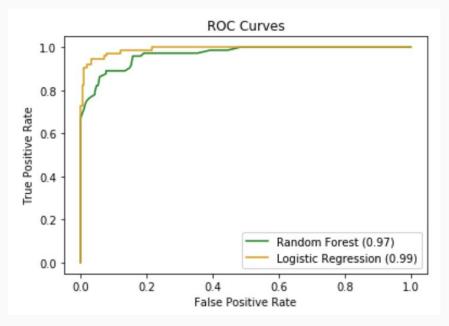
- Oversampled minority class to balance our target variable
- Classifiers generally performed well on training set
- Logistic regression showed the best accuracy and area under ROC
- Expected ensemble methods to perform better

	accuracy	roc_auc
name		
LogisticRegression	0.967071	0.997560
RandomForestClassifier	0.976761	0.996773
MultinomialNB	0.821548	0.898377



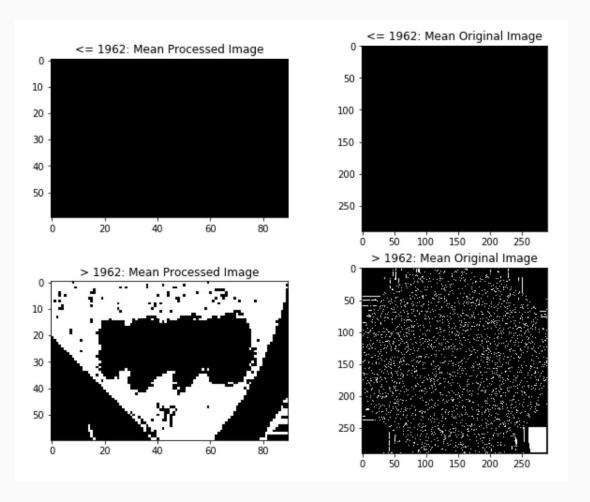
Classifier Comparison - Training Data





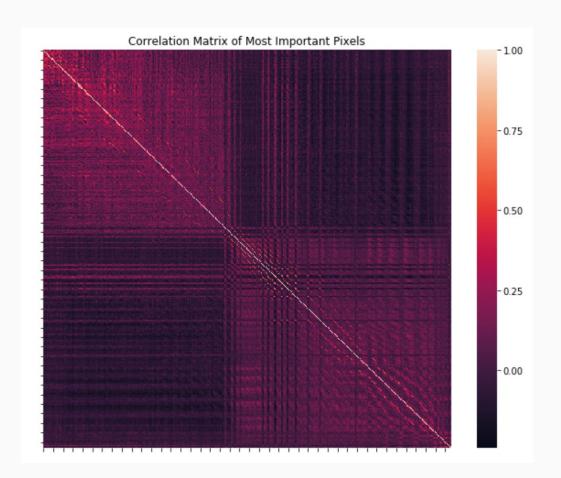
Target Variable

- Coin years after 1962 are brighter than before 1962
- This effect is preserved through preprocessing



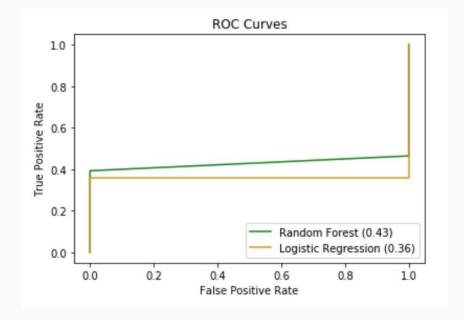
Collinearity

- Looking at the 500 most important pixels, we see some patterns.
- Most of the important pixels are on the border of the year on the coin



Classifier Comparison - Test Data

- Significant drop in performance when applying to test data
- Suspected causes:
 - Small amount of test data
 - Image preprocessing could be more robust



Next Steps

- Look at individual digits
- Better image preprocessing
- Collect more data and attempt to classify all years
- Predict coin condition (good, fair, excellent, etc.)
- Explore Deep Learning algorithms

