

# Rock Paper Scissors



A image classification model created by Michael Shepherd and Connor Rogers

# Hypothesis:

Using image classification, it is possible to determine the hand gestures a user is showing to play a game of rock, paper, scissors against a computer.

Rock =



# Data Set: Rock Paper Scissors

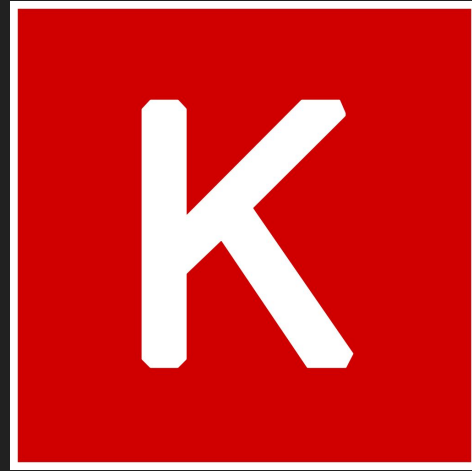
- 700+ images for each sign.
- Images are of resolution 300 x 200.
- Images were taken with consistent Lighting and Background.
- Dataset taken from a similar project, although our model uses the full resolution images, not 1/2 scale versions

(<https://www.kaggle.com/drgfreeman/rockpaperscissors>)



# Classification Model:

- Keras/TensorFlow Neural Network.
- Uses 2D Convolutions to exaggerate the feature from the background.
- 2D Pooling is used to downsample output for subsequent operations.
- Dropouts provide randomizations, to reduce overfit scenarios.



# Model Training:

- Augmentations of each image were created (Flips, Rotations, Etc.) using Keras' `ImageDataGenerator.flow_from_directory()` to avoid having to manually load and augment each image.
- 80% of data was used for training, 20% was used for validation steps during training.
- Training terminated when model met 98% accuracy during validation to avoid overfitting.

```
generated_data = ImageDataGenerator(rescale=1. / 255, rotation_range=20, vertical_flip=True,  
horizontal_flip=True, shear_range=0.2, fill_mode='wrap', validation_split=0.2)  
  
train_data = generated_data.flow_from_directory(base_dir, target_size=image_size,  
class_mode='categorical', subset='training', shuffle=True)  
  
val_data = generated_data.flow_from_directory(base_dir, target_size=image_size, class_mode='categorical',  
subset='validation', shuffle=True)
```

# Deployment: Server

- Deployed as a Flask application running Gunicorn, a WSGI HTTP server for UNIX, hosted by NginX.
- OpenCV for handling input from client.
- Image is processed by Flask server, classified using the pre-trained Keras model, and the result (with a random choice of sign by the CPU) is sent back to the client.

```
def predict(img_source):  
    img = expand_dims(img_source, axis=0)  
    classes = model.predict(img, batch_size=10)[0]  
    if classes[0] > classes[1] and classes[0] > classes[2]:  
        result = "paper"  
    elif classes[1] > classes[0] and classes[1] > classes[2]:  
        result = "rock"  
    elif classes[2] > classes[0] and classes[2] > classes[1]:  
        result = "scissors"  
    return result  
  
def decide_winner(player_choice):  
    cpu_choice = choice(['rock', 'paper', 'scissors'])  
    if player_choice == cpu_choice:  
        return [player_choice, cpu_choice, "tied"]  
    elif player_choice == "scissors" and cpu_choice == "rock":  
        return [player_choice, cpu_choice, "lost"]  
    elif player_choice == "rock" and cpu_choice == "paper":  
        return [player_choice, cpu_choice, "lost"]  
    elif player_choice == "paper" and cpu_choice == "scissors":  
        return [player_choice, cpu_choice, "lost"]  
    else:  
        return [player_choice, cpu_choice, "won"]
```

# Development: Client

- Live viewfinder created using JavaScript's `mediaDevices.getUserMedia()` to work with most modern browsers
- Captured image is converted from HTML canvas to dataURL in base64 format, then sent to Flask backend using ajax
- Response is received from server using ajax and HTML is updated using JavaScript's `document.getElementById()`

Capture Function to get image from user webcam:

```
async function capture() {  
  ...  
  var canvas = document.getElementById('c');  
  canvas.width = video.videoWidth;  
  canvas.height = video.videoHeight;  
  canvas.getContext('2d').drawImage(video, 0, 0,  
video.videoWidth, video.videoHeight);  
  var dataURL =  
  canvas.toDataURL().split(';base64,')[1];  
}
```

Ajax request used to send image to server and receive response:

```
$.ajax({  
  type: "POST",  
  url: "/get_image",  
  data: {imageBase64: dataURL},  
  success: function (res) {  
    document.getElementById('message').innerHTML = res ;  
  },  
  error: function (error) { console.log(error) }  
})
```

## Conclusion:

- Using Keras, it was possible to create a Neural Network to determine with accuracy what hand sign (rock, paper, or scissors) a user is showing their camera
- Using this and a Flask based web server, a user can play rock, paper, scissors with a computer simply by visiting a webpage
- Deployed as a desktop and mobile-friendly front end  
(not working on IOS devices for unknown reasons)

Visit <https://www.CodeSmith.link> if you would like to play!