

# Autonomous Computer-Vision-Based Human-Following Robot

Akshay Gupta, Nick Gloria

{akshaygu, ngloria}@stanford.edu

## Introduction

- For “golf caddy” product: autonomous golf bag carrier
- Computer vision directs robot to follow human, navigate terrain
- Enhances golfer experience, assists elderly or disabled golfers
- Robot has single camera, onboard processor
- Terrain classification problem: classifying image into traversable/untraversable terrain (avoiding green, sand traps, etc.)
- Target-following problem: identify and track target, maintaining distance

## Data and Features

- Dataset collected via onboard camera (Jetson SoC)
- Training and test data sets selected from images taken during product testing
- Supervised data provided by hand-classified, color-coded images

## Models

Both unsupervised and supervised methods were implemented to this classification problem.

### Target Tracking: Convolutional Neural Network - MobileNet SSD

- Using pre-trained neural network, identified following target with bounding box

### Image Classification: K-means-based Segment Detection

- K means clustering on a test images resulting in color based clusters
- Clusters picked for appropriate terrain areas, which generates segments based on norm distance in the color space

### Image Classification: Felzenszwalb Superpixel Segmentation + Pooling

- SciKit based image segmenter, results in boundary generation based on pixel wise difference both locally, and globally after clustering
- Draws boundaries after generating clusters segmenting into large regions
- The pixel values are averaged in each cluster and repainted onto image

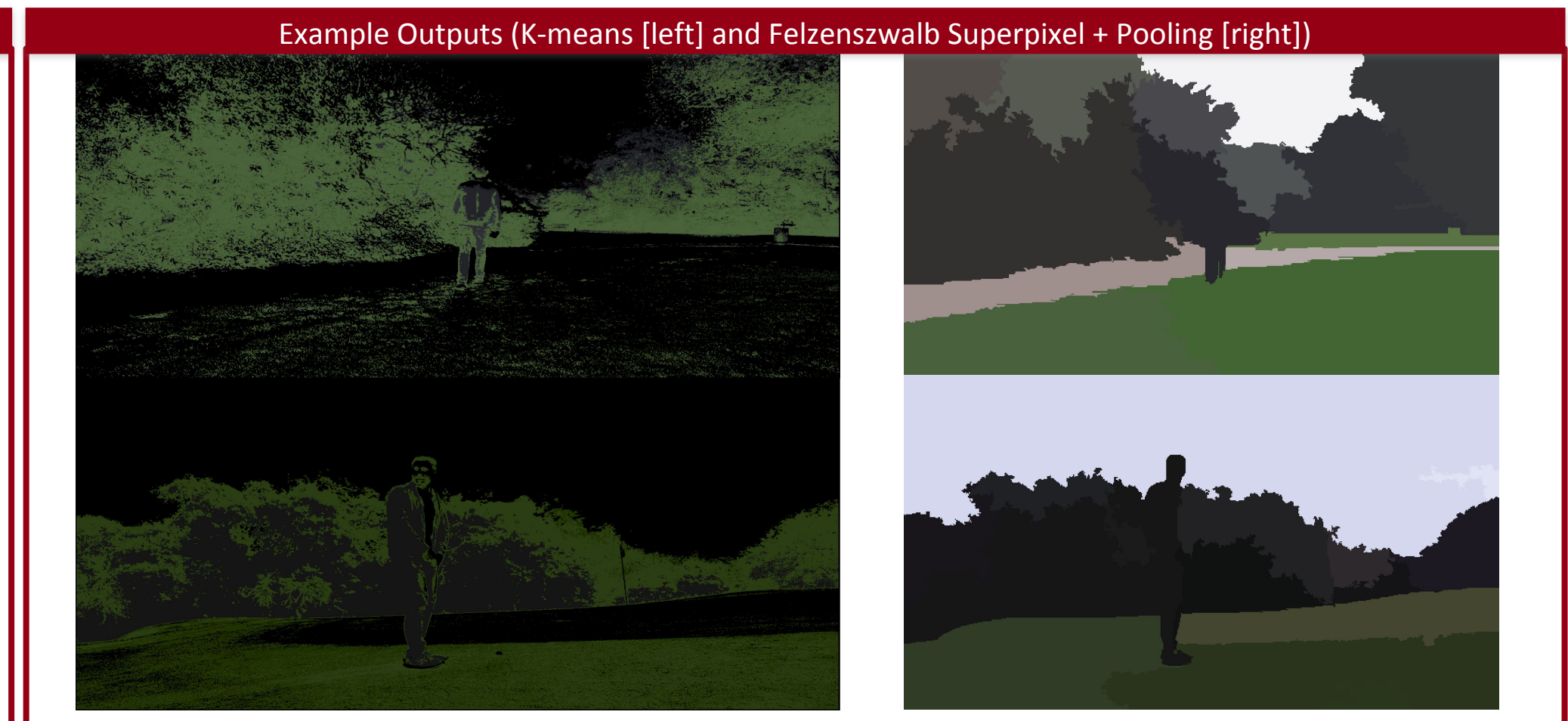
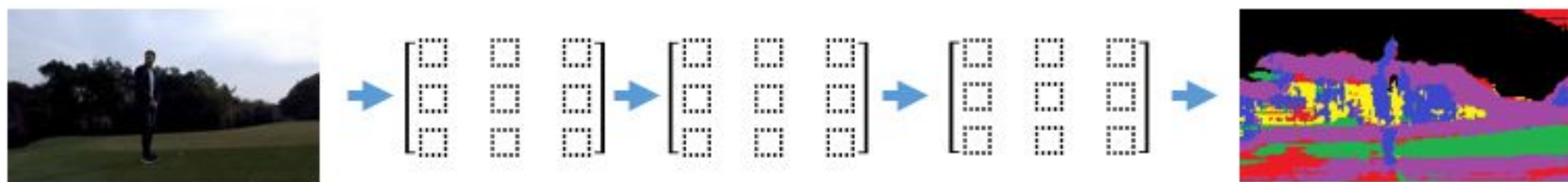
### Image Classification: Convolutional Layer

- Square pixel stencil converts each pixel to a vector of pixels
- Linear combination of neighboring pixel vectors gives access to gradient

### Image Classification: Softmax Regression

- Uses key-painted images for training
- Classifies each pixel into one category
- Classes: green, traversable grass, bunker, target, path, etc.

Convolution      Softmax  
Regression



## Results and Discussion

The results show that with given data, the algorithms implemented can roughly classify the robot images, with some shortcomings and pitfalls.

### Convolution + Softmax Regression

- Rough accuracy with noise, limited by single-layer classification
- Suffers from low number of hand-classified images for training

### Image Classification: K-means-based Segment Detection

- Pixelated images generated with area segmentation and limited closed curves
- Cluster selection needs tuning and automation to increase efficiency

### Image Classification: Felzenszwalb Superpixel Segmentation + Pooling

- Good clustering but fails to capture all boundaries, also generates more clusters than required

### Target Tracking: Convolutional Neural Network - MobileNet SSD

- High accuracy and precise detection, unable to detect smaller or thin objects
- Needs to be fine tuned on golf course setting, including equipment and obstacles

## Future Steps

- Generation of larger dataset on golf course images for training the networks
- Use edge-detection algorithm to implement pooling layer (i.e. use most common classification to classify identified segment)
- Extension of single classification to multiple classification layers (neural network)
- Devise efficient means of classifying images (e.g. augmentation by edge detector, stamping) to allow for expansion of training set

## References

- [1] Felzenszwalb, P.F. & Huttenlocher, D.P. International Journal of Computer Vision (2004) 59: 167. <https://doi.org/10.1023/B:VISI.0000022288.19776.77>
- [2] Howard, et al. “MobileNets: Efficient Convolutional Neural Networks for Mobile Vision Applications.” [Astro-Ph/0005112] A Determination of the Hubble Constant from Cepheid Distances and a Model of the Local Peculiar Velocity Field, American Physical Society, 17 Apr. 2017. [arxiv.org/abs/1704.04861](https://arxiv.org/abs/1704.04861).