

# Object Position Prediction from human hand motion

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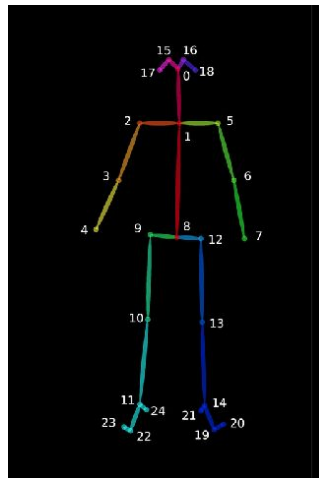


# Introduction

- Prediction of the object location based on the motion of the human right wrist
- Five objects in known locations
- No object localization or object detection algorithms

# Data Collection

- RGB-D camera for human monitoring
- OpenPose for human wrist 2D pose estimation
- Predetermined and known object locations
- Predetermined and known starting position of the human wrist





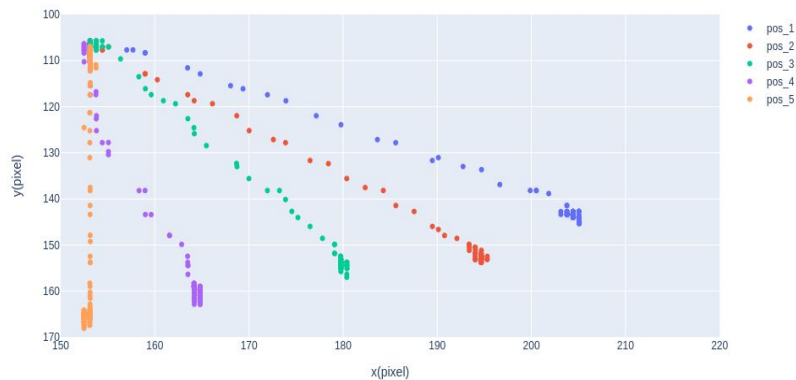
# Data Collection

- RGB frames collected @ 60Hz (RGB-D sensor frequency)
- OpenPose data
  - ~22Hz (OpenPose frequency based on utilized GPU)
  - 60Hz
- Human motion representation
  - 2D OpenPose Pixels
  - 3D Cartesian Coordinates (OpenPose 2D pixels + PointCloud depth information)

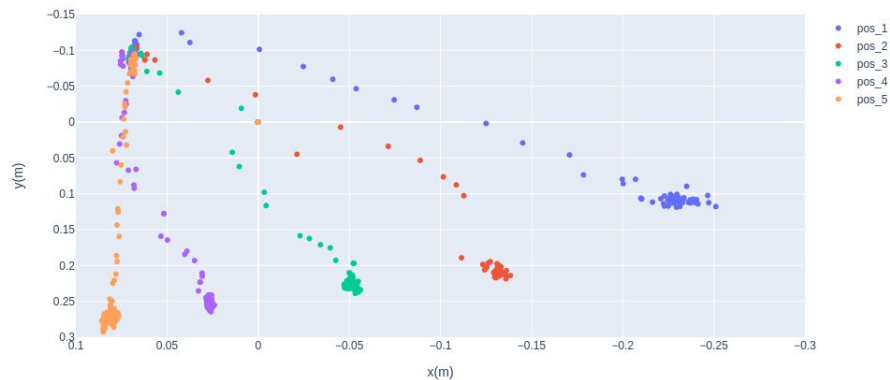


# Data Collection

OpenPose pixels



Keypoints



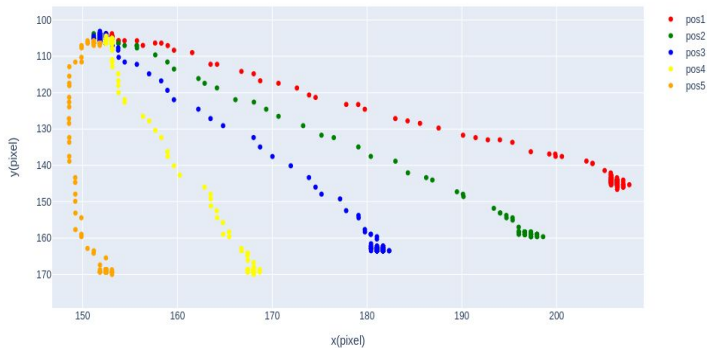
- OpenPose 2D pixels due to less noise



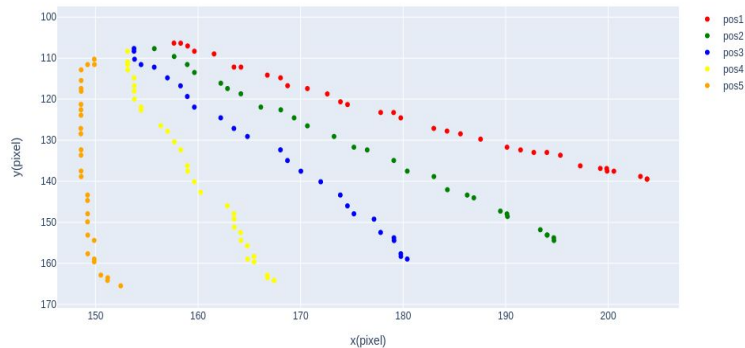
# Data Preprocessing

- NaN values removal
- Outlier removal
- Static pixels removal

Raw Openpose pixels



Clean Openpose pixels





# Feature Extraction

- Features: x, y OpenPose pixels
- Real-time behaviour
  - Pixels are obtained over time
  - Window of increasing length encapsulating available pixels at time t
    - $t=t_0 \Rightarrow [x_0, y_0], t=t_1 \Rightarrow [x_0, y_0, x_1, y_1], \dots$
  - Class labels:  $y_i=\{1, 2, 3, 4, 5\}$





# Training/Validation

- Naive Bayes, Decision Tree, SVM
- Training/Testing
  - Dataset @ 22Hz (OpenPose frequency)
  - Dataset @ 60Hz
- KFold split (n=10)
- SVM
  - kernel={rbf, linear}
  - C={0.1, 1, 10}

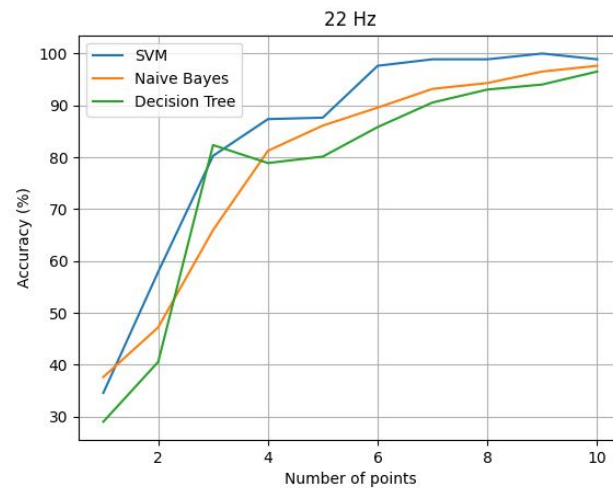
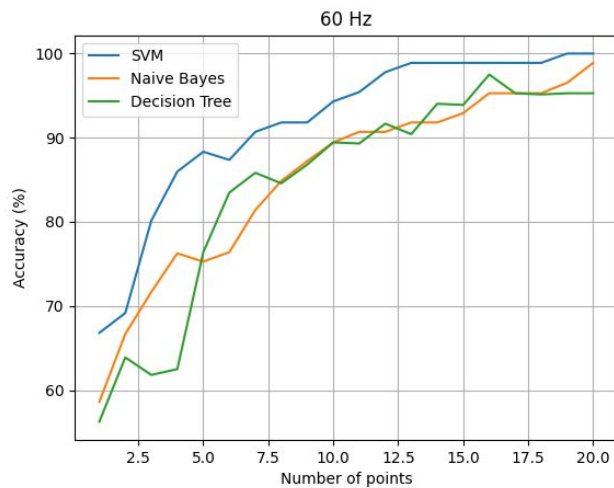


# Training/Validation

- Metrics
  - Accuracy: How good and at which time do the algorithms predict the human intended goal?
  - Confusion Matrix: In misclassification cases, what did the algorithm predict?



# Training/Validation





# Training/Validation

- SVM
  - Kernel: rbf
  - $C=1$
- SVM performs better than Naive Bayes and Decision Tree classifiers

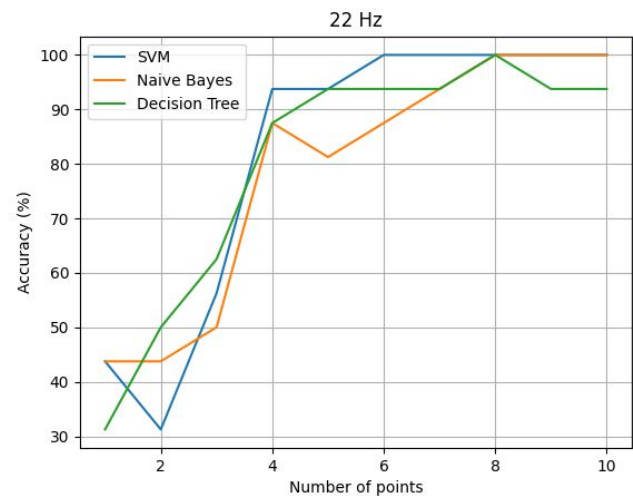
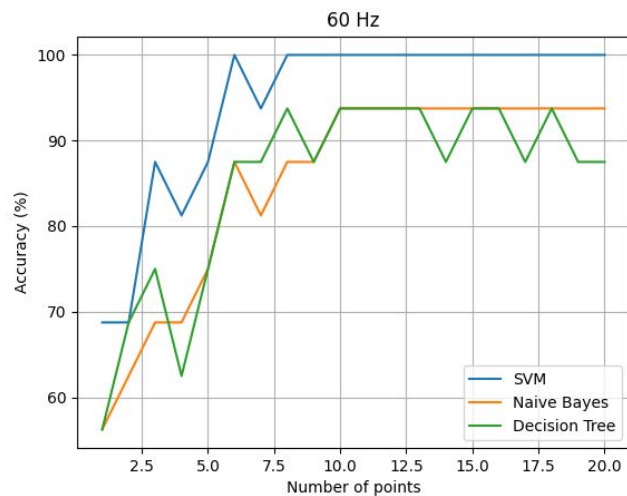


# Testing

- Training using the entire training-validation datasets
- Store the models for each classifier
- Test in testing dataset
- Metrics: same as in the training/validation phase



# Testing



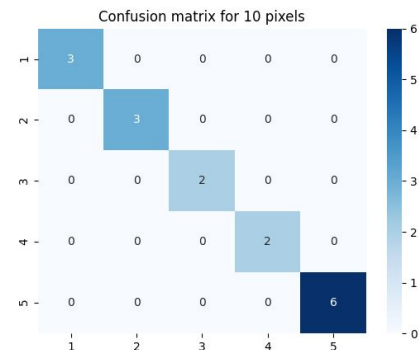
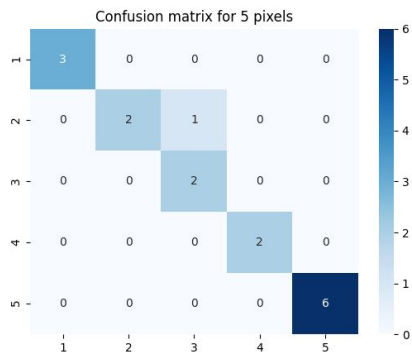
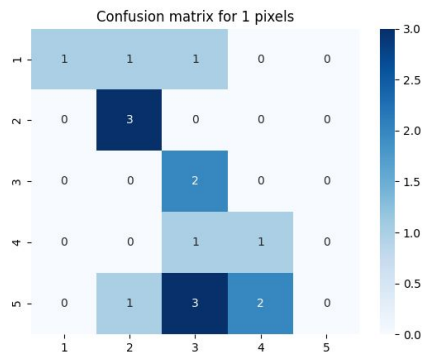


# Testing

- SVM with linear kernel and  $C=1$  produces the most accurate predictions
  - >90% accuracy in the first 188ms (22Hz) and 100ms(60Hz)
- Naive Bayes and Decision Trees
  - 85% accuracy in the first 188ms (22Hz) and 100ms (60Hz)
- Having available the pixels at 60Hz instead of 22Hz results in approximately 0.8-1.2 times faster prediction.



# Testing







# Limitations

- Objects clearly separable
- Human initial position relatively far from the objects
- Complex human motions e.g. obstacles in the workspace