Abnormal Gait Diagnosis with wearable device

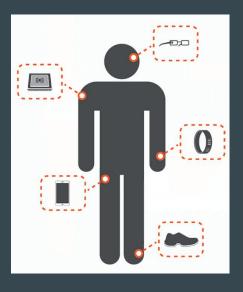
Zhaoxi Zhang Mechanical Engineering

Outline:

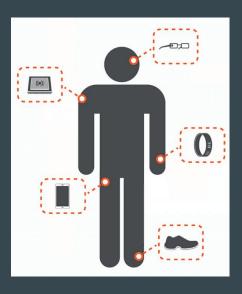
- Background
 - o Data from wearable device
 - Importance of gait abnormalities
- Motivation
 - Telemedicine
 - o Potential
- Project Components
 - o Hardware
 - o Algorithm
 - Future work: Interface

• Importance of gait

- Importance of gait
- Popularity of wearable devices



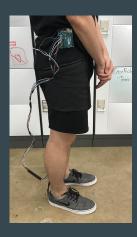
- Importance of gait
- Popularity of wearable device
- Willingness to share data

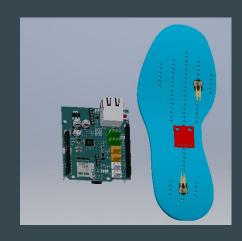


- Importance of gait
- Popularity of wearable device
- Willingness to share data
- Project potential influence
 - Shorten diagnosis time
 - Cost saving

Hardware

- 9 DOF IMU and Force Sensor on Shoe insole
 - Acceleration on x,y,z axes
 - Angular acceleration around x,y,z axes
 - Voltage to represent force

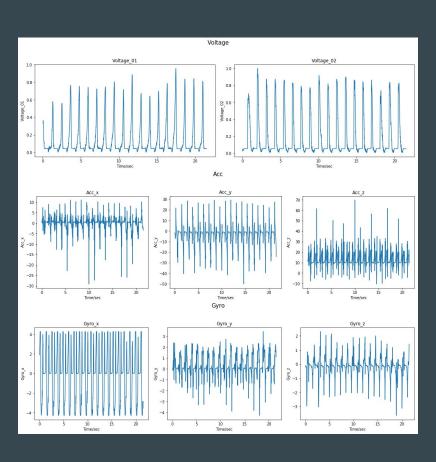








Hardware



- Four Categories
 - Tendem Gait
 - o Inward
 - Outward
 - o Normal



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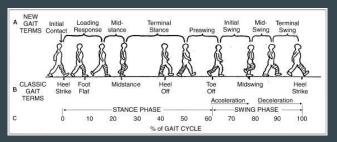


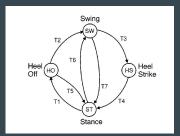
- Four Categories
 - Tendem Gait
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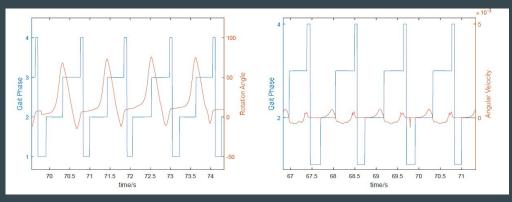


- Gait Phase State Machine
- Feature Engineering and classification algorithm
- Long Short Term Memory Recurrent Neural Network for classification

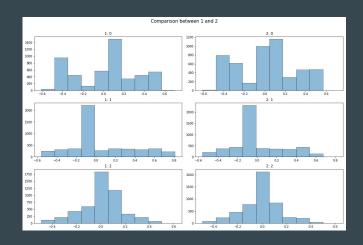
- State Machine Approach:
 - o Four Gait Phases
 - Stance
 - Heel Off
 - Swing
 - Heel Strike
 - Normal Gait Transition:
 - T1: ST to HO
 - T2: HO to SW
 - T3: SW to HS
 - T4: HS to ST
 - Non-walking Transition:
 - T5: HO to ST
 - Abnormal Gait Transition:
 - T6: ST to SW
 - T7: SW to ST

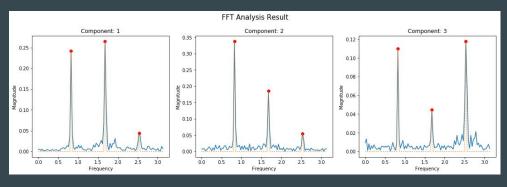




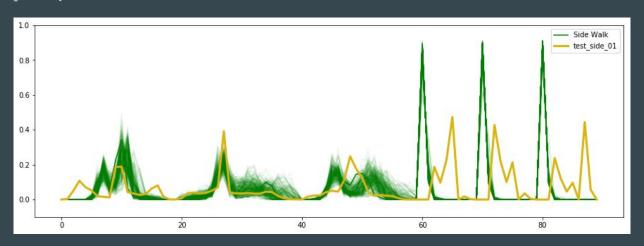


- Feature Engineering
 - O Dimension Reduction: Principal Component Analysis (PCA)
 - o Distribution Feature: Histogram
 - Frequency Feature: FFT

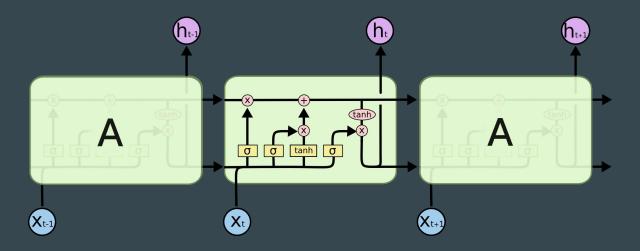




- Feature Engineering
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Long Short-Term Memory RNN



Outward

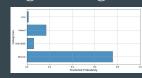


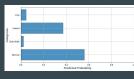


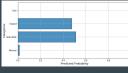


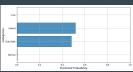


- Feature Engineering
- + Logistic Regression

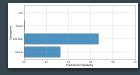


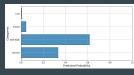


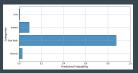


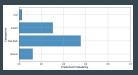


- Feature Engineering
- + SVM Platt Scaling

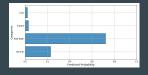


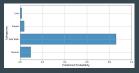


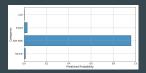


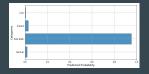


LSTM RNN









Interface

- Use instructions with caveats
- Interpretable diagnosis result
 - o diagnosis results
 - Example videos of certain abnormality to help understanding
 - Potential solution
 - Caveats

Summary

- Importance of gait abnormality
- Hardware and experiement
- Algorithm and experiment
- Futhure work: interface

Collaboration

• The gait phase state machine approach is implemented collaboratively with YuanKai Zhu.

Reference

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Github: https://github.com/vacous/GaitDiagnosisML

Thank You