

# Mobile App Based in Home Parkinson Disease Management

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# Outline

- What is the problem (PD)?
- Significance of wearables in PD space
- App Design
  - Design decisions
  - Symptoms
  - In-home monitoring
  - Sensor data collection
  - App activities and views
- Machine learning for prediction
- What is supervised learning?
- What is self-supervised learning?
- Results

# Do you know about him?



<https://slate.com/technology/2016/06/what-caused-muhammad-alis-parkinsons-disease-its-nearly-impossible-to-say.html>

# Parkinson Disease (PD)

- Progressive disorder (Age>50)
- Affects the nervous system and the parts of the body controlled by the nerve

## Common Symptoms:

- Tremors
- Gait & Balance Problem  
Freezing of Gait: A sudden and temporary inability to initiate or continue walking, as if their feet are glued to the ground.

- No cure



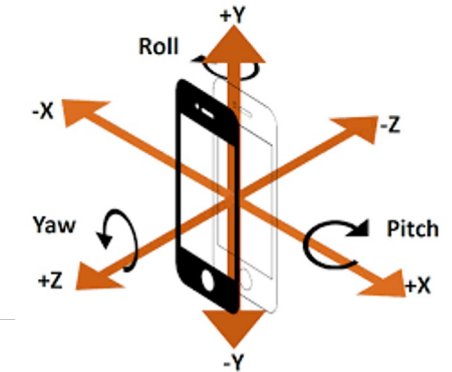
# Wearables in PD

## IMU Sensors (6D)

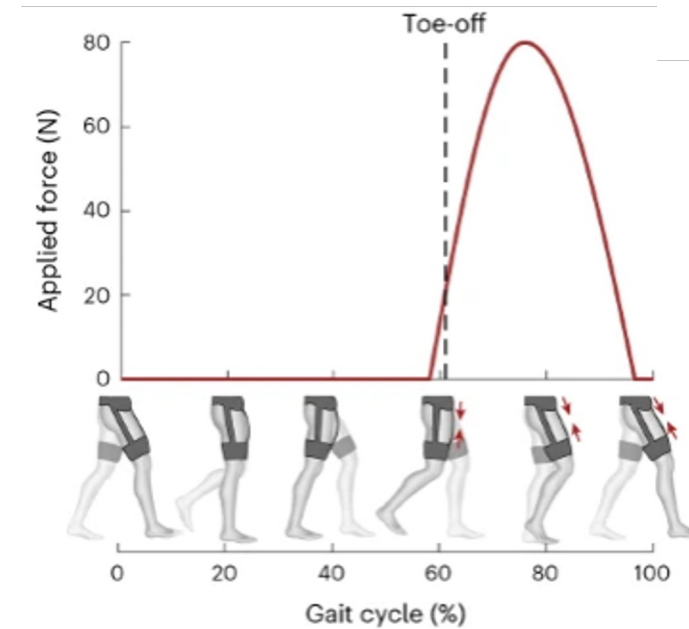
- Accelerometer (3D): Acceleration
- Gyroscope (3D): Angular velocity
- Extract Different Features (Velocity, and gait pattern, detect physiological biomarkers)
- Early PD Detection
- Detect deterioration for known PD patients



1 DEVICE

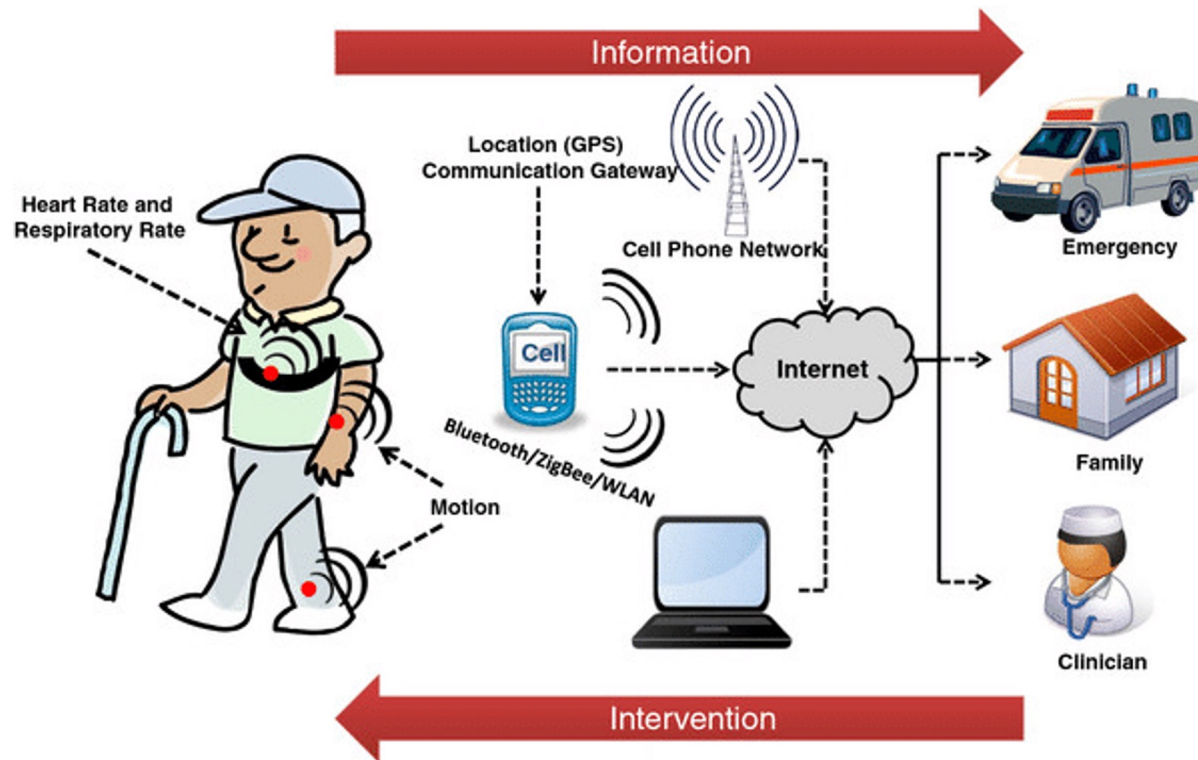


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# Wearables in PD

Store the data collected by wearables into a cloud service



# App Design

There are 2 features to be implemented.

1. **Capture Sensor Data** - Accelerometer, Gyroscope, Heart rate, step counter.
2. **Implement PD Hauser Diary**

	ON	ON with dyskinesia	OFF	Asleep
<b>06:00–06:30</b>				X
06:30–07:00				X
<b>07:00–07:30</b>			X	
07:30–08:00			X	
<b>08:00–08:30</b>	X			
08:30–09:00	X			
<b>09:00–09:30</b>	X			
09:30–10:00	X			
<b>10:00–10:30</b>	X			
10:30–11:00			X	
<b>11:00–11:30</b>			X	
11:30–12:00			X	

# App Design

The app captures the data related to hauser dairy.  
A wireframe design of app -

Time	Asleep	ON	On With troublesome dyskinesia	On Without troublesome dyskinesia	OFF
6-6:30AM	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input checked="" type="checkbox"/>



# App Design

- The sensor data collection is made by smartwatch app.
- The app collects sensor data of accelerometer, gyroscope, heart rate, step counter.
- The composite data is then to synced cloud every 10 minutes if the watch is worn.
- There also a feature on the app to indicate medication is taken.
- To optimize the battery usage data is collected at 50Hz frequency.

# What is ML?

## Traditional Programming

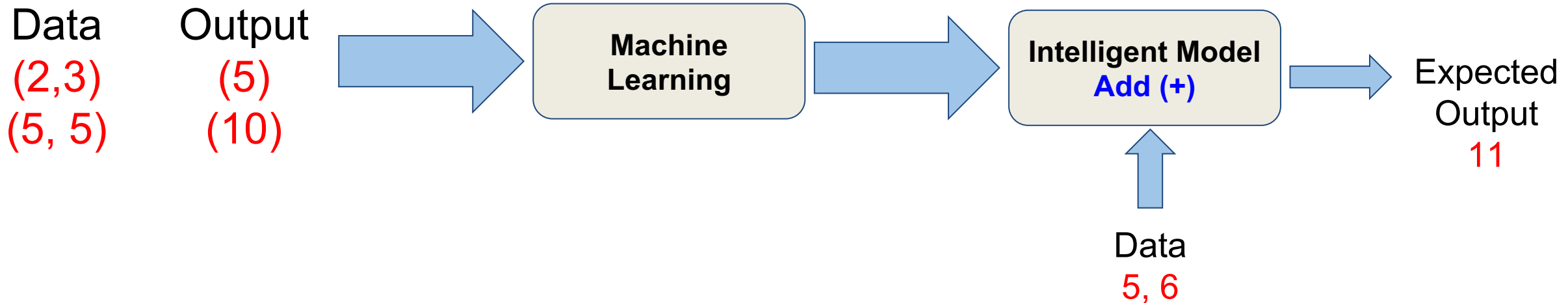
- Takes inputs and generates outputs based on the program
- Programs are fixed unless they are updated manually



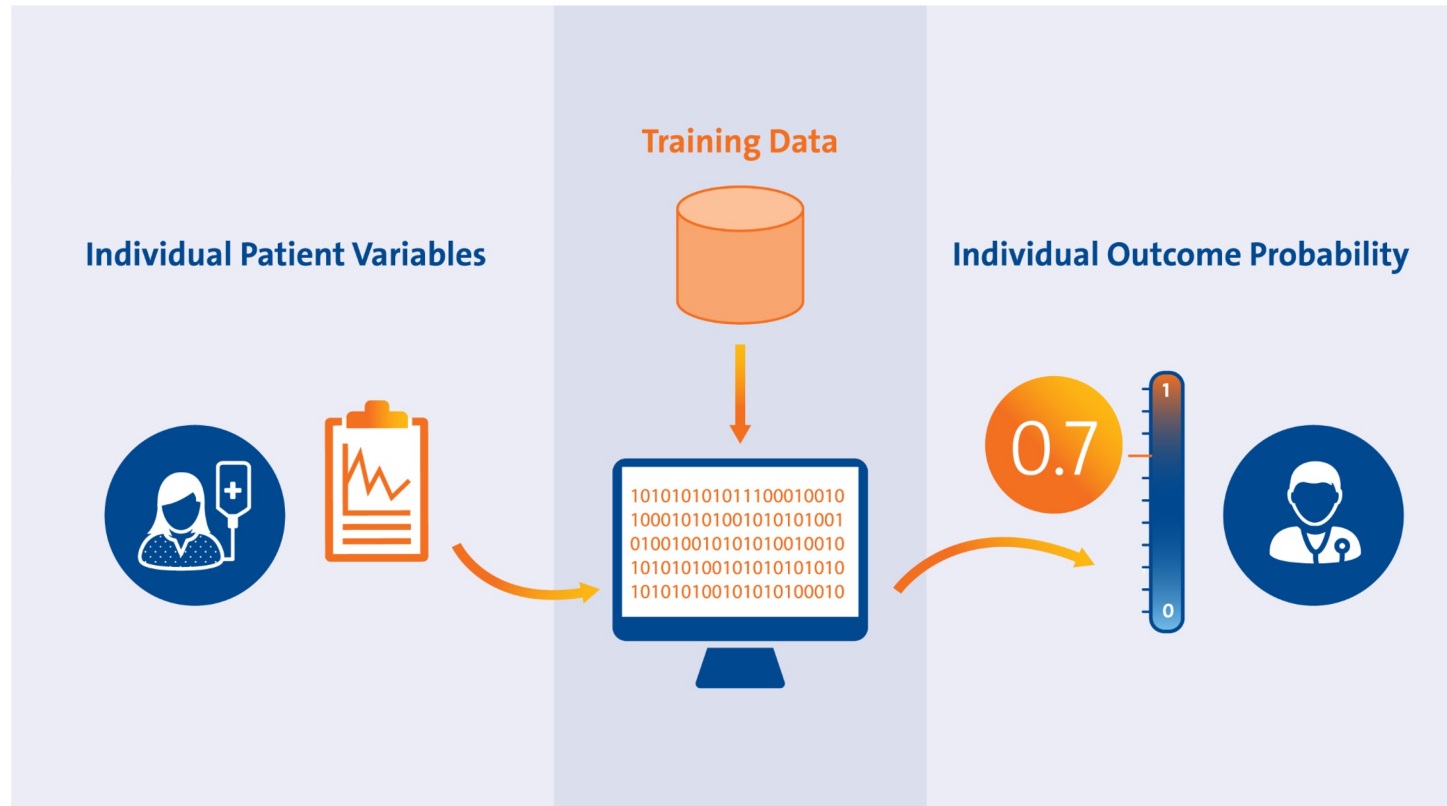
# What is ML?

## Machine Learning

- Takes data and expected output as input
- Learns the data patterns/representations without any type of explicit programming
- Continuously evolve into better models

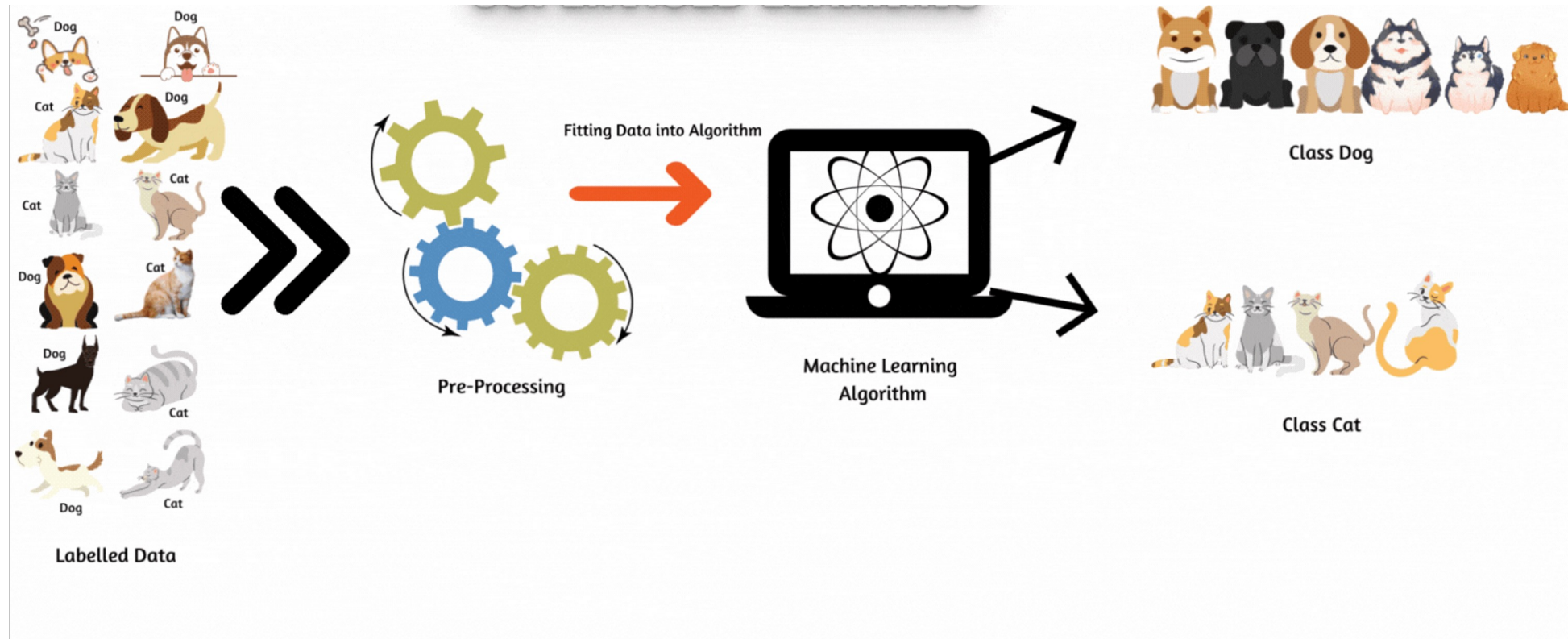


# ML in Prediction



# Supervised Learning

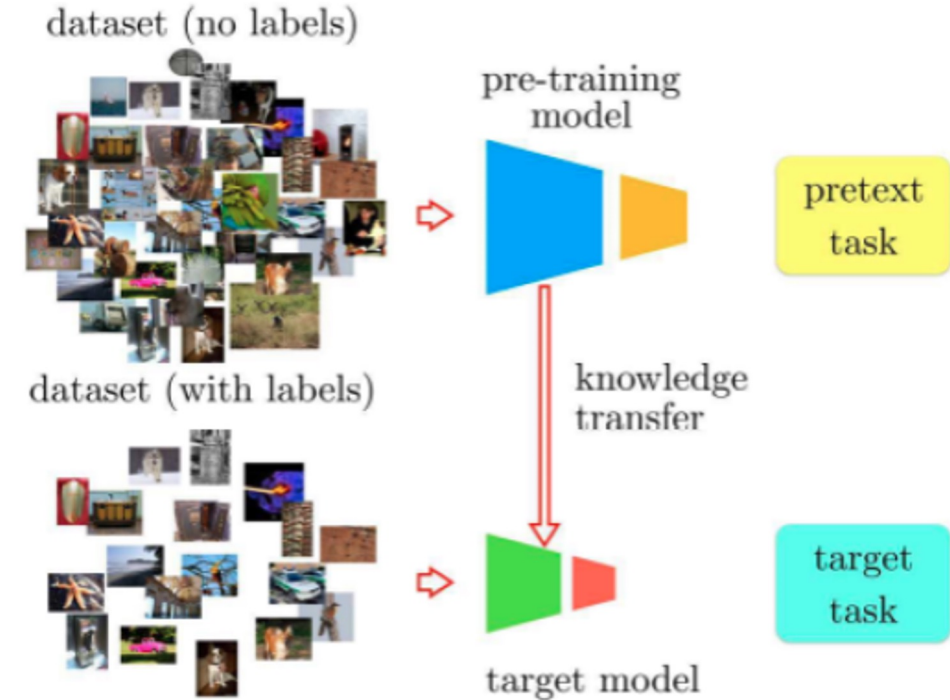
Train & Test both have labels (ground truth)



# Self Supervised

Train the models in 2 steps

1. **Pretraining**: Uses **unlabeled data** and learns some interesting features/patterns
2. **Finetune**:
  - a. Uses the previous learned knowledge
  - b. **Retrains** the learned knowledge using **few labelled data** to solve a specific problem

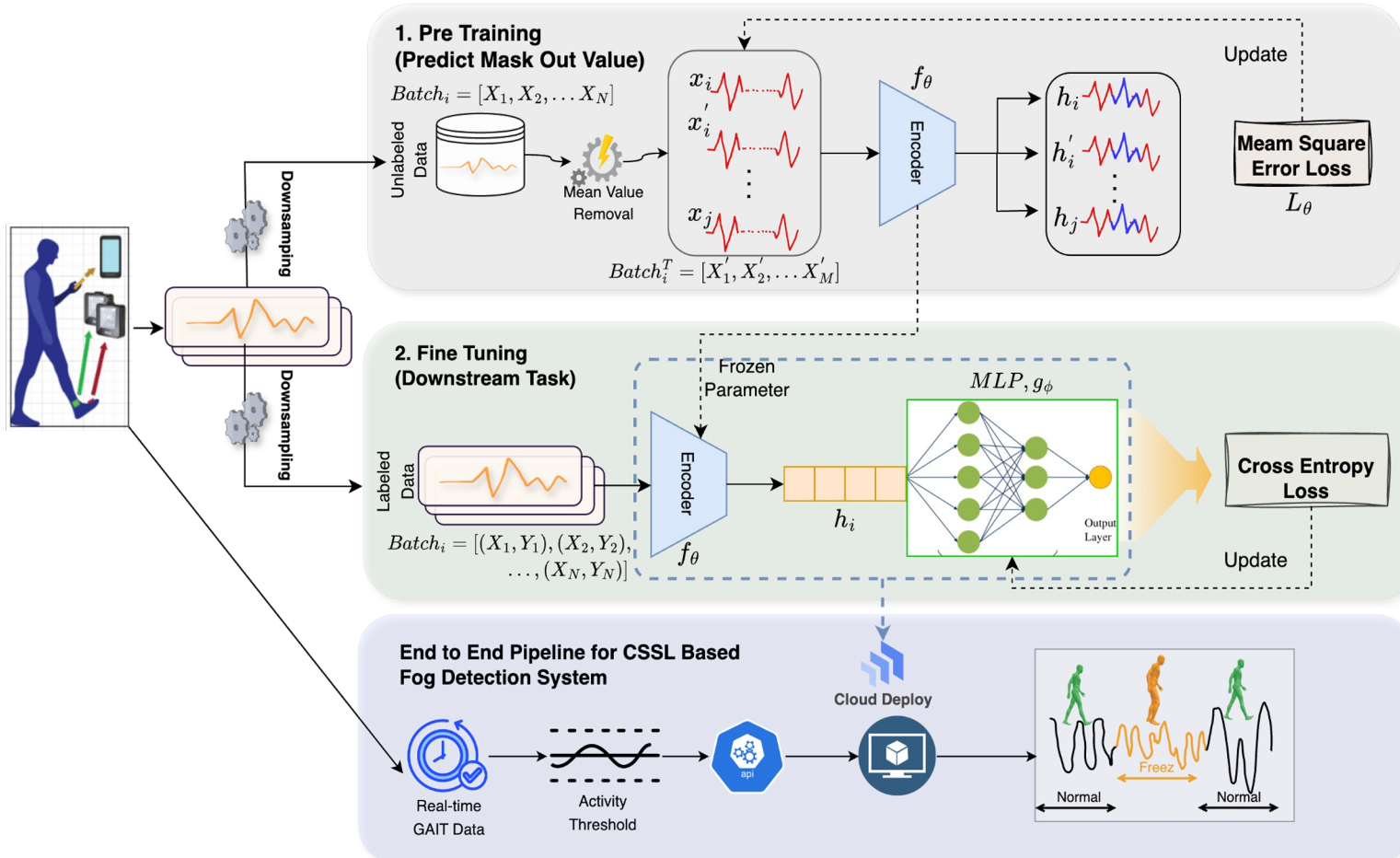


# *On Going PD Research*

**DEAP-ML: Data-Efficient, Adaptive, and Power-Aware Machine Learning for Parkinson's Long-Term Home Monitoring**

# DEAP-ML: Data-Efficient, Adaptive, and Power-Aware Machine Learning for Parkinson's Long-Term Monitoring

- A single IMU sensor → Reducing patient burdens
- Patient-independent Model
- Cost-effective opportunistic AI model
- Uncontrolled home environment for long time remote monitoring.
- Optimized for extended battery life



a. *Pretext Task*: Raw signals  $X \in \mathbb{R}^{N \times T}$   
 Masked window,  $\hat{X} = \text{mask}(X, m)$

Trained Pretext Model,  $f_{\theta}$

$$L_{\theta} = \frac{1}{N_m} \sum_{j=1}^{N_m} (\hat{x}_j - x_{p(j)})^2$$

b. *Down Stream Task*: MLP on top of pretext model

$$\hat{Y} = h_{\theta, \phi}(X) = g_{\phi}(f_{\theta}(X))$$

$$L_{ce}(\theta, \phi) = -\frac{1}{N} \sum_{i=1}^N [y_i \log(\hat{y}_i) + (1 - y_i) \log(1 - \hat{y}_i)]$$



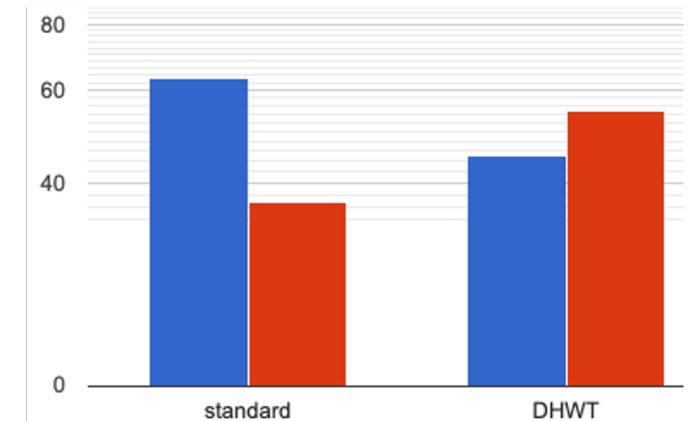
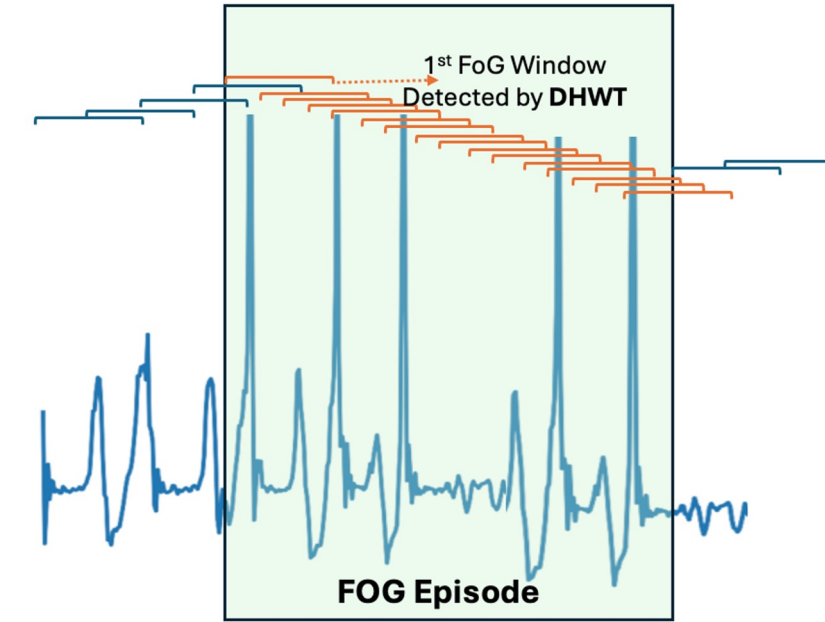
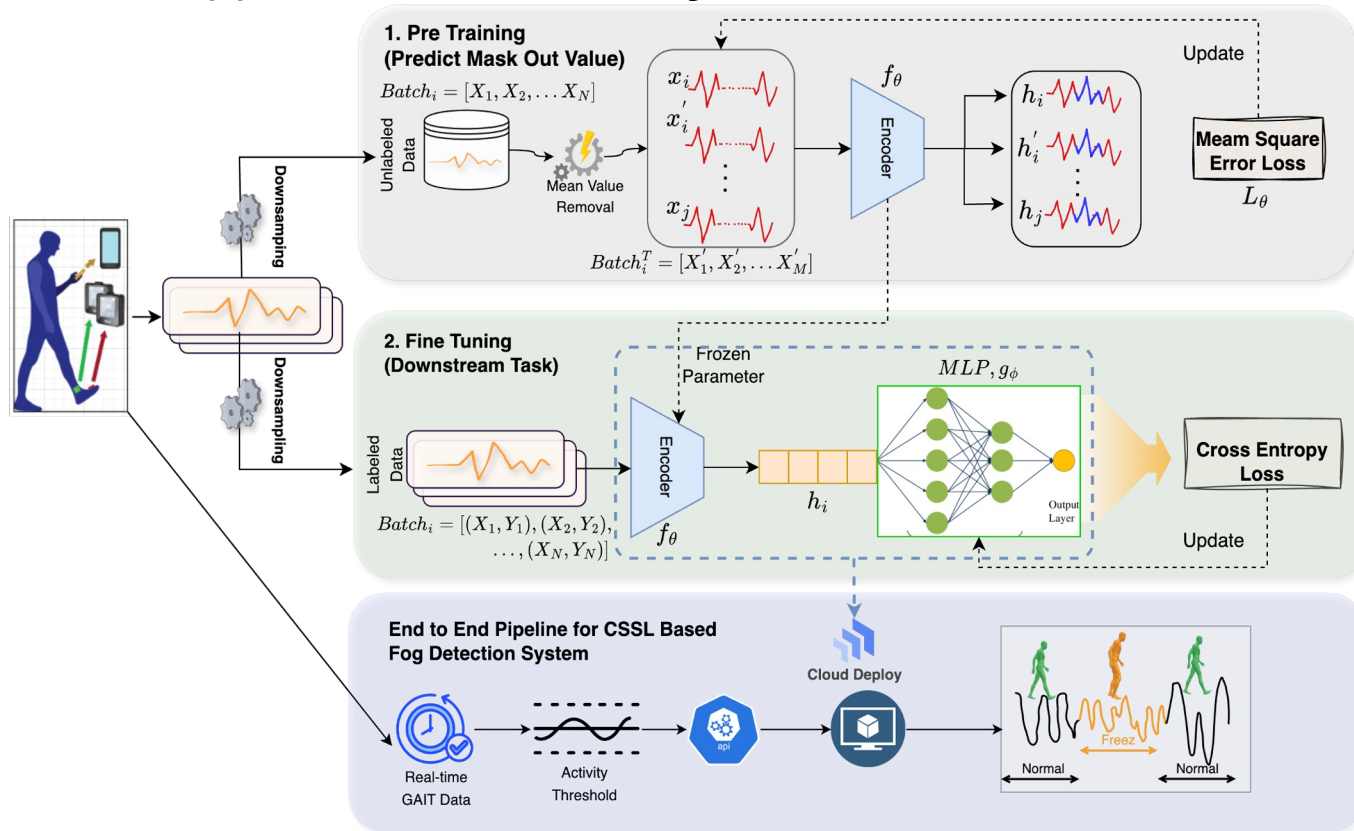
# DEAP-ML: Data-Efficient, Adaptive, and Power-Aware Machine Learning for Parkinson's Long-Term Monitoring (Contd)

## Differential Hopping Windowing Technique (DHWT)

- Handle the imbalance dataset with low FoG samples

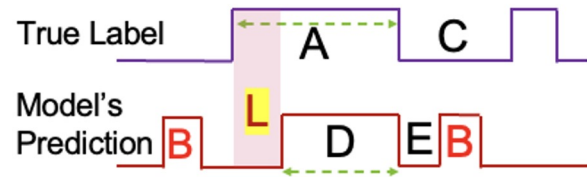
## Minimal Preprocessing (Removing Mean-value from each window)

- Applied to **trainset only** to mimic real-world



# Results

- Using small amount data we got promising results
- Long FoG episodes are easily detectable

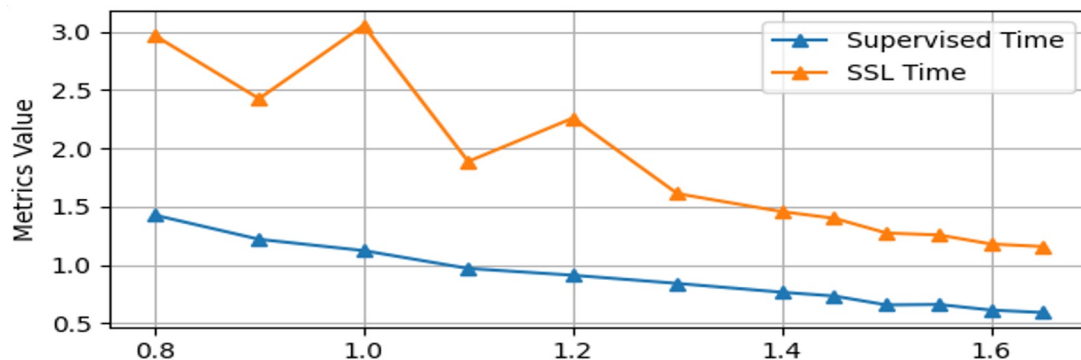
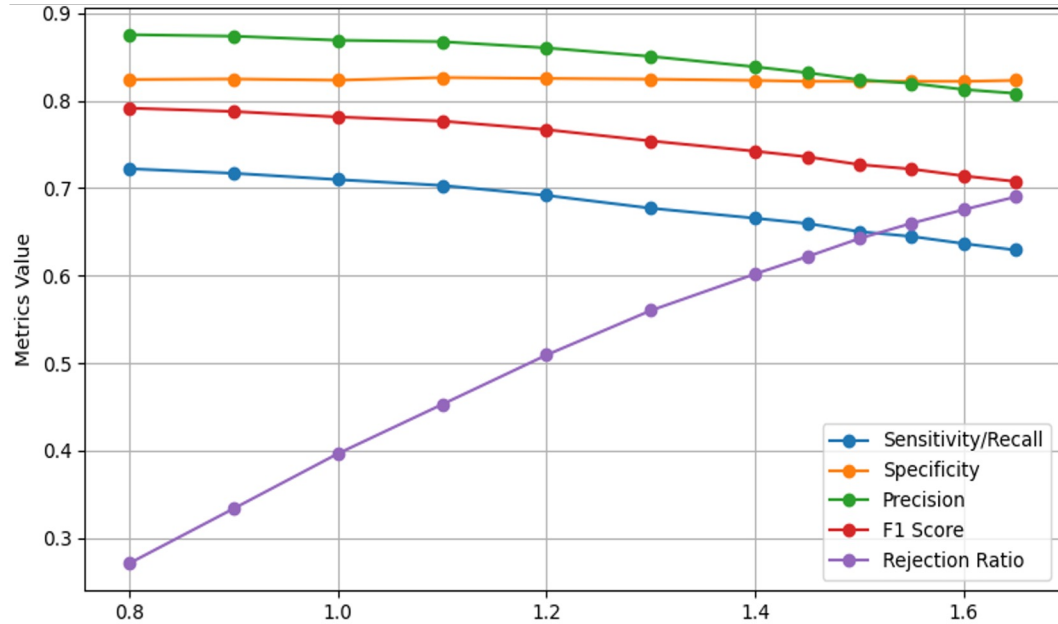


A: FoG Episode                      L: Latency  
 B: False Positive                      C: Normal Episode  
 D: Detected FoG Episode (TP)  
 E: Detected Normal Episode (TN)  
**\*\*False alarm triggers just before/after a TFE**

	Precision	F1	Accuracy	FoG Episodes	FoG windows, (Recall/ Sensitivity)
SSL	86.5%	79.7%	76.6%	<b>86.6%</b>	74.2
SUP	90.1%	86%	83.1%	<b>86.6%</b>	82

Episode Time (sec)	Avg Latency	Detected Episodes	Detected Windows (%)
< 5 sec	0.18s	88%	73%
6-12 sec	0.83s	86.21%	67%
<b>&gt;13 sec</b>	<b>2.35s</b>	<b>97.3%</b>	<b>82.4%</b>

# Results



$$\text{magnitude, } m = \sqrt{x^2 + y^2 + z^2}$$

- Set a magnitude Threshold ( $\alpha$ ) to discard the inactive windows.

If  $m < \alpha$  : discard the current window as it is inactive (not related to gait activities)

- It will extend the battery life
- Model will be lightweight → easy to deploy into a mobile wearable devices

# Future Work

- **Suggest Effective Medication Plan**
  - Medication time
  - Medication dose
- **Assess PD Severity**
- **Detect atypical and typical PD**

# Thank You!

**Lab Website: <https://ghasemzadeh.com/>**

## Members

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