

# Detection of Parkinson's Disease Using Automated Tunable Q Wavelet Transform (A-TQWT) Technique with EEG Signals

Digital Signal Processing

*Erasmus Mundus Joint Master Degree in Medical Imaging and Applications*

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# Parkinson's Disease

- **Parkinson's disease** is a brain disorder that causes unintended or uncontrollable movements, such as shaking, stiffness, and difficulty with balance and coordination.
- It affects the **nervous system** and the parts of the body controlled by the nerves.
- It is a **neurodegenerative disorder** that affects predominately the dopamine-producing (“dopaminergic”) neurons in a specific area of the brain called **substantia nigra**.

# Symptoms of Parkinson's Disease

- Involuntary shaking of particular parts of the body such as hands, arms, legs, jaw, or head (tremor).
- Slow movement.
- Depression and anxiety.
- Balance problems.
- Loss of sense of smell (anosmia).
- Memory problems.

# Diagnosis and Detection of the Parkinson's Disease

- **Interviewing and questionnaire method**
  - Interviewing and questionnaire methods by trained neurologists are time-consuming, burdensome and prone to errors.
- **Voice based methods**
  - The performance of voice-based methods are limited and can be altered intentionally.

# Diagnosis and detection of the Parkinson's Disease

- **Neuroimaging methods**

- Computed Tomography, PET Imaging, and Functional MRI.
- It require additional recordings, time-consuming, costly and radio-active as compared to signal processing techniques.

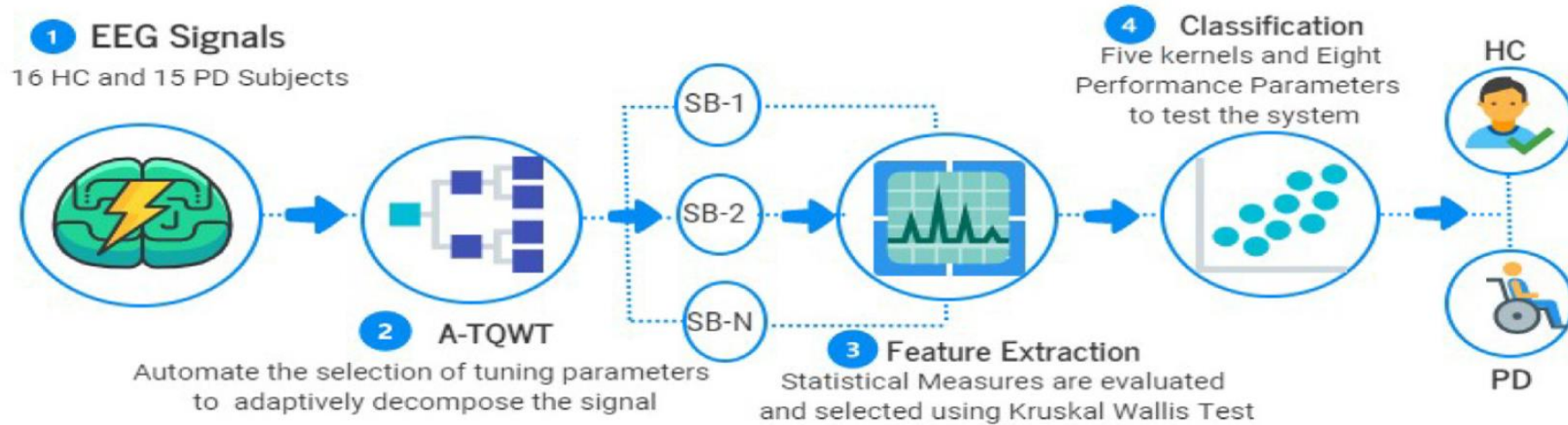
- **Signal Processing methods**

- Electrocardiography(ECG), Electromyography (EMG), and Electroencephalography (EEG).
- The detection of PD using ECG and EMG signals has been accomplished but **offers limited success** in terms of classification accuracy.
- EEG signals have **yielded high performance** in the detection of PD even with its low amplitude.

# Techniques for Identifying PD Using EEG Signal

- Fast Fourier Transform (FFT).
- Analysis of Variance (ANOVA) with Fast Fourier Transform (FFT).
- Discrete Wavelet Transform (DWT).
- Discrete Wavelet Transform (DWT) and Fast Fourier Transform (FFT).
- **Tunable Q-Wavelet Transform (TQWT).**

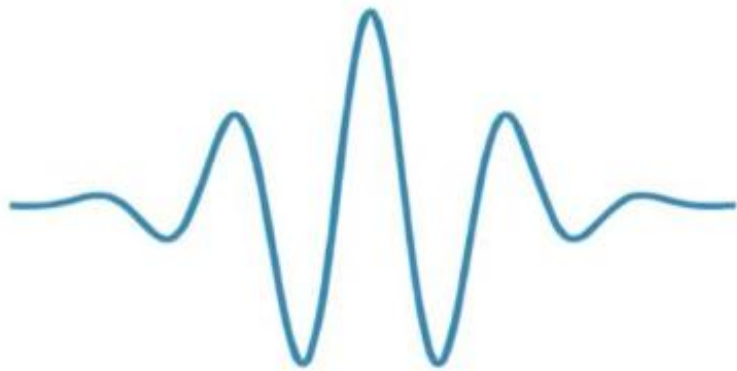
# Proposed Method



Flow Chart of the Proposed Method.

# Wavelet Transform (WT)

- A **Wavelet** is a **wave-like oscillation** that is localized in time.
- It has average value of zero.



Wavelet

## Wavelet Equation

$$\psi_{a,b}(t) = \frac{1}{\sqrt{a}} \psi\left(\frac{t-b}{a}\right) \quad a, b \in \mathbf{R}$$

where,

$a$  = Scaling (Dilation) Factor.

$b$  = Shifting (Translation) Factor.



# Wavelet Transform (WT)



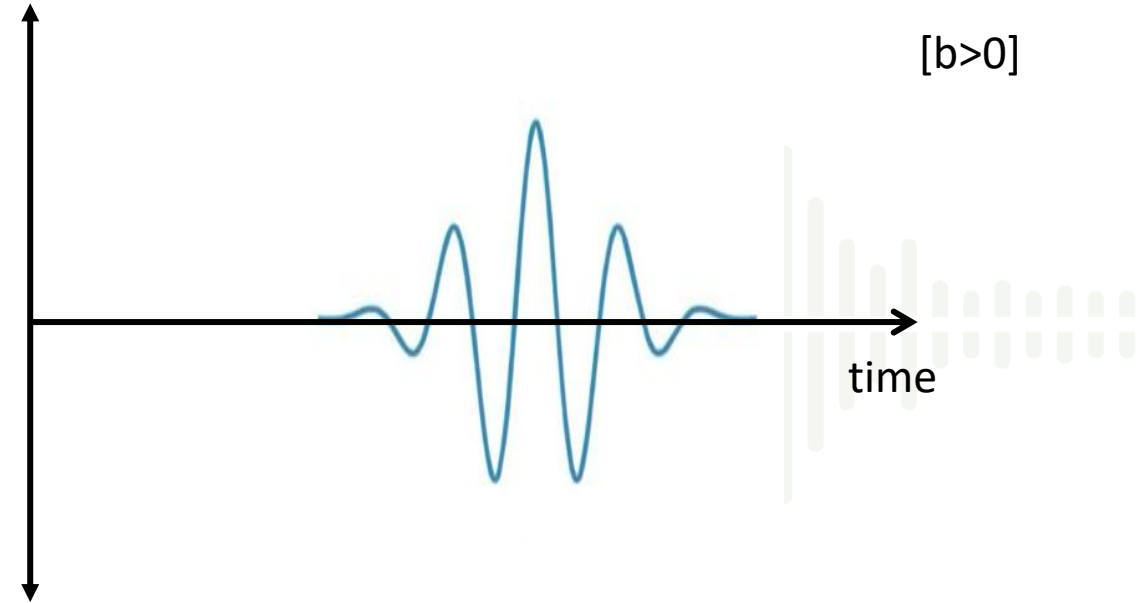
$[a > 1]$

Compressed Wavelet for Extracting  
High Frequency Component



$[0 < a < 1]$

Stretched Wavelet for Extracting  
Low Frequency Component

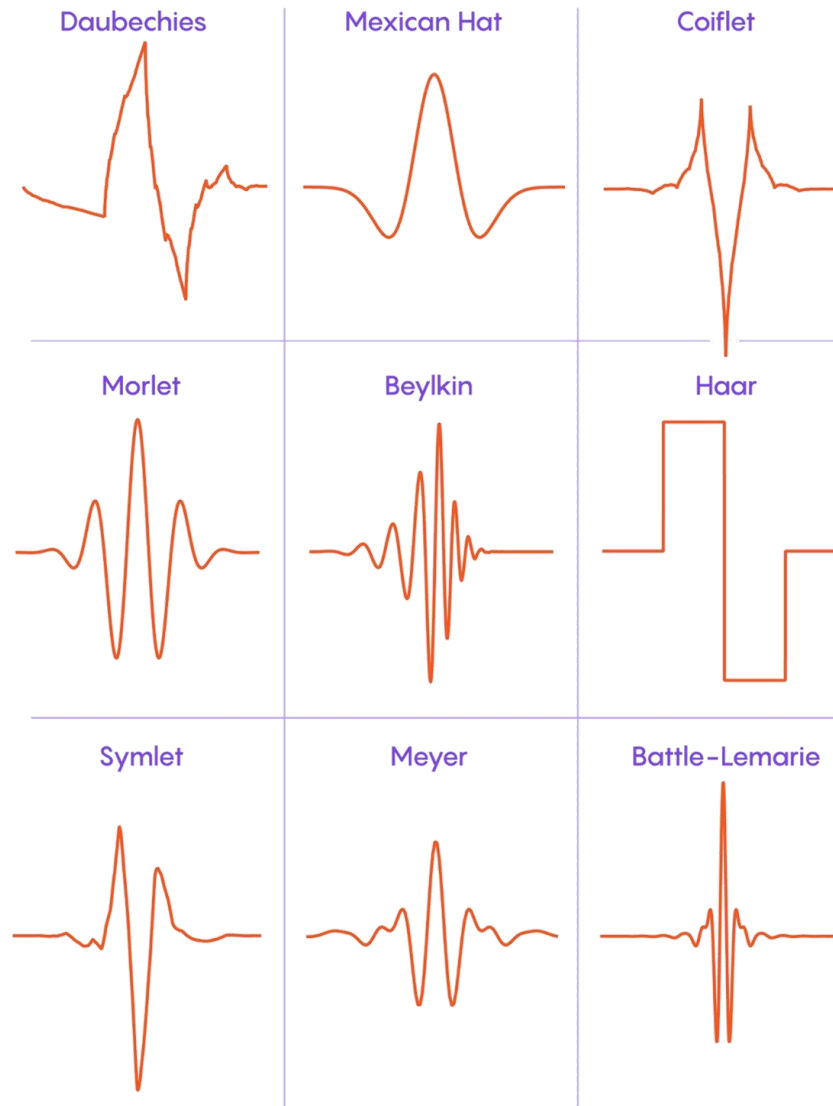


$[b > 0]$

Right Shifted Wavelet

# Wavelet Transform (WT)

- Different types of wavelets



# Wavelet Transform (WT)

- The **wavelet transform** is a mathematical technique which can **decompose a signal into multiple lower resolution levels** by **controlling the scaling and shifting factors** of a single wavelet function (also called “mother wavelet”).
- **Equation for Continuous Wavelet Transform**

$$T(a, b) = \frac{1}{\sqrt{a}} \int_{-\infty}^{\infty} \mathbf{x(t)} \psi^* \left( \frac{t-b}{a} \right) dt$$

where,

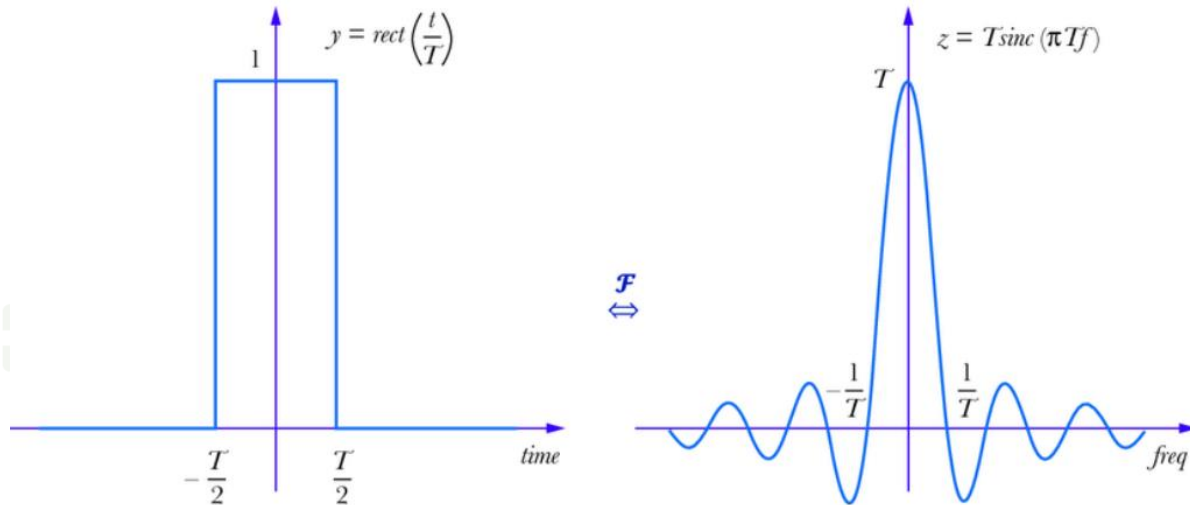
x(t) = Input/Main Signal.

a = Scaling Parameter.

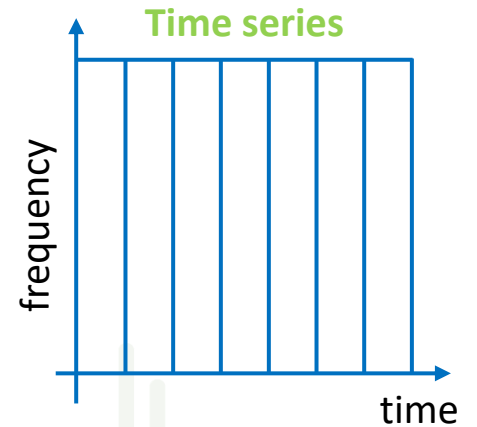
b = Shifting Parameter.

# Wavelet Transform (WT) vs Fourier Transform (FT)

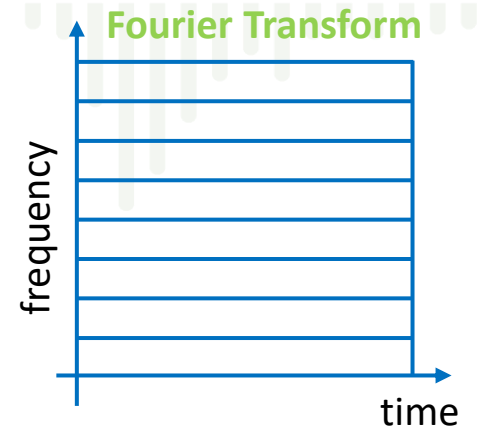
- Fourier Transform of a Rectangular Pulse



- Fourier Transform (FT) provides global frequency information.
- Either time or frequency can be known at a specific point.



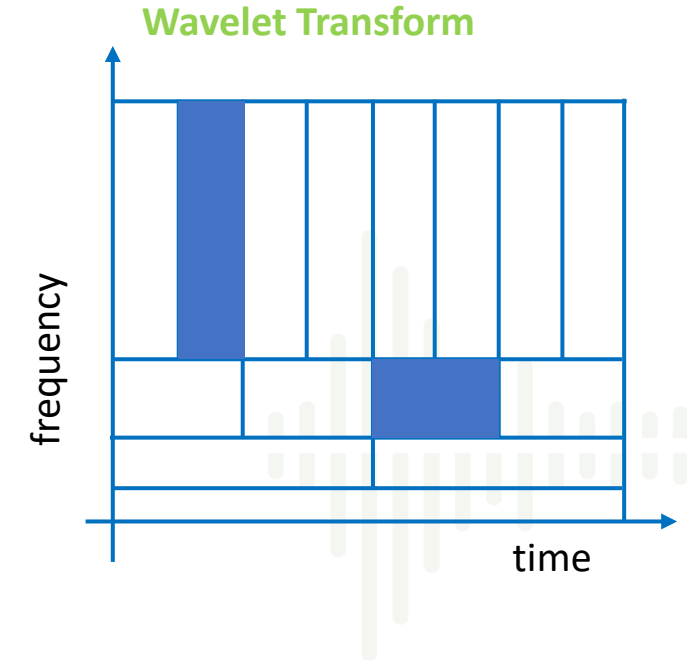
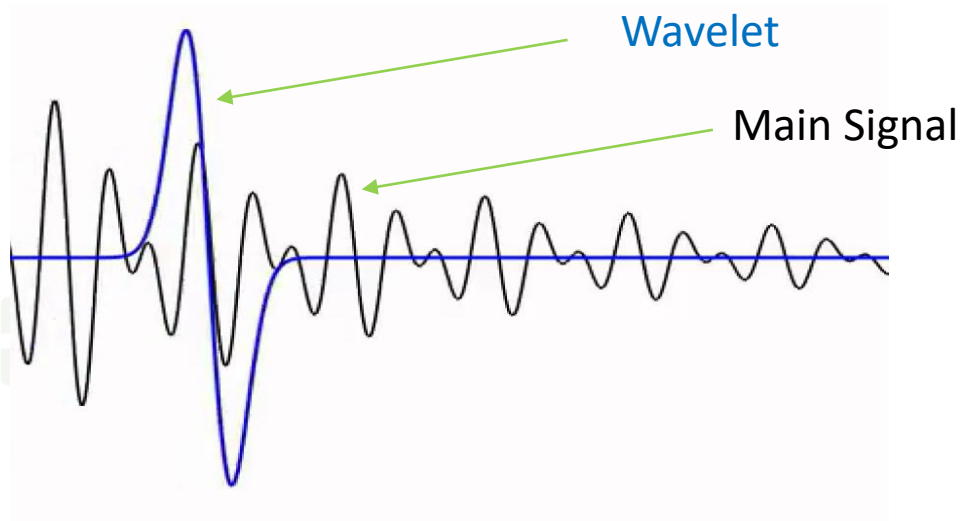
High time resolution  
Zero frequency resolution



High frequency resolution  
Zero time resolution

# Wavelet Transform (WT) vs Fourier Transform (FT)

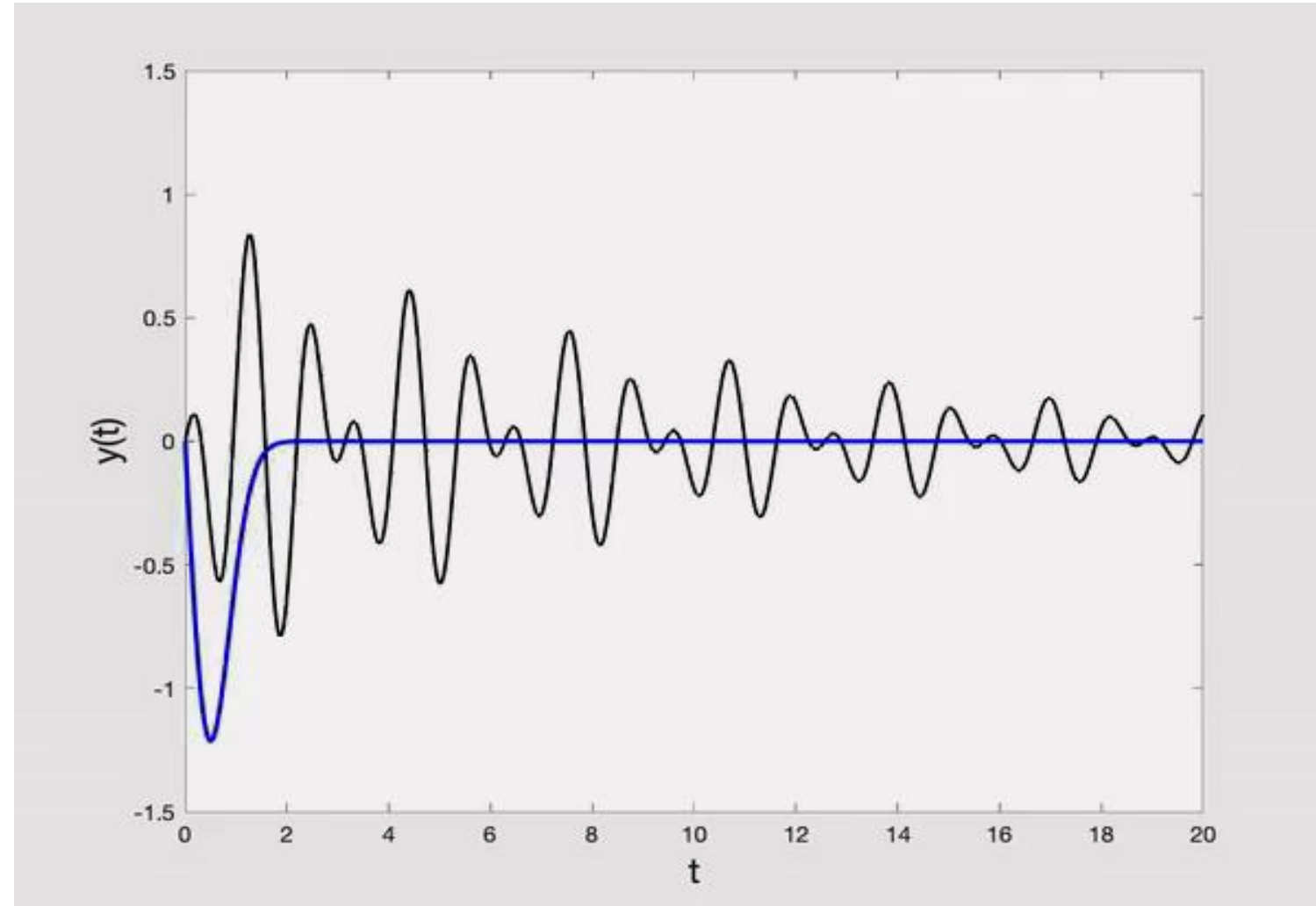
- Wavelet Transform



- Wavelet Transform (WT) provides specific frequency information.
- Both time or frequency can be known at a specific point.
- For small frequency values a high resolution in the frequency domain, low resolution in the time- domain.
- For large frequency values a low resolution in the frequency domain, high resolution in the time domain.

# Feature Capturing Using Wavelet Transform

- It is like a **convolution operation**.
- Input signal is fixed while **wavelet is moving** from left to right.



# Discrete Wavelet Transforms (DWT)

- For Continuous Wavelet Transform, scaling (a) and translation (b) are selected for every possible point which create a huge amount of data.
- Therefore, we can only consider the discrete points of scaling (a) and translation (b) parameters for small amount of data.

## ➤ Discrete Wavelet

$$\psi_{m,n}(k) = a_0^{-\frac{m}{2}} \psi(a_0^{-m}(k - nb_0 a_0^m))$$

$$\begin{cases} a = a_0^m & [a_0 > 1, m = \text{width}] \\ b = nb_0 a_0^m & [b_0 > 0] \end{cases}$$

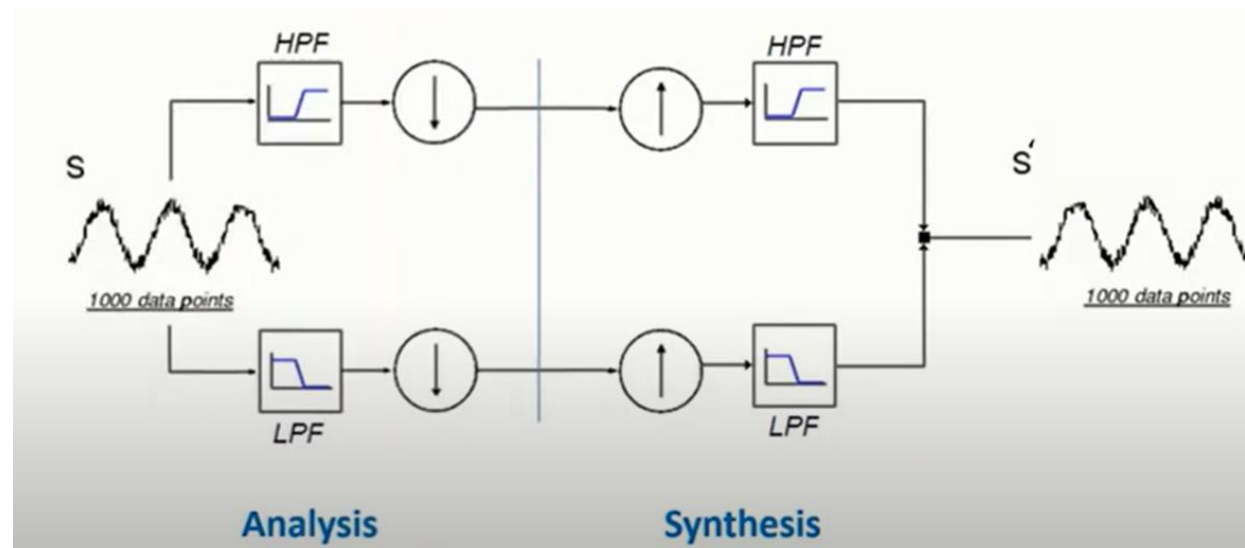
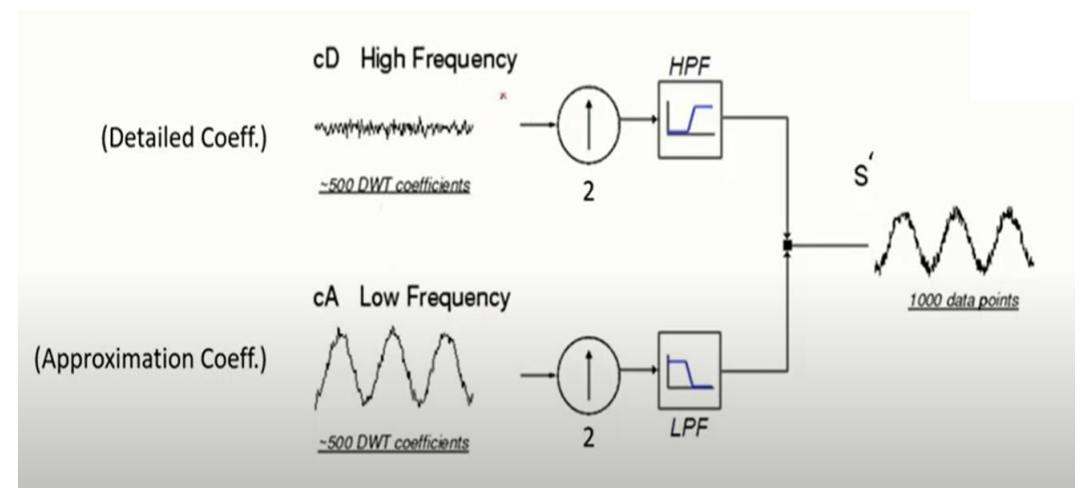
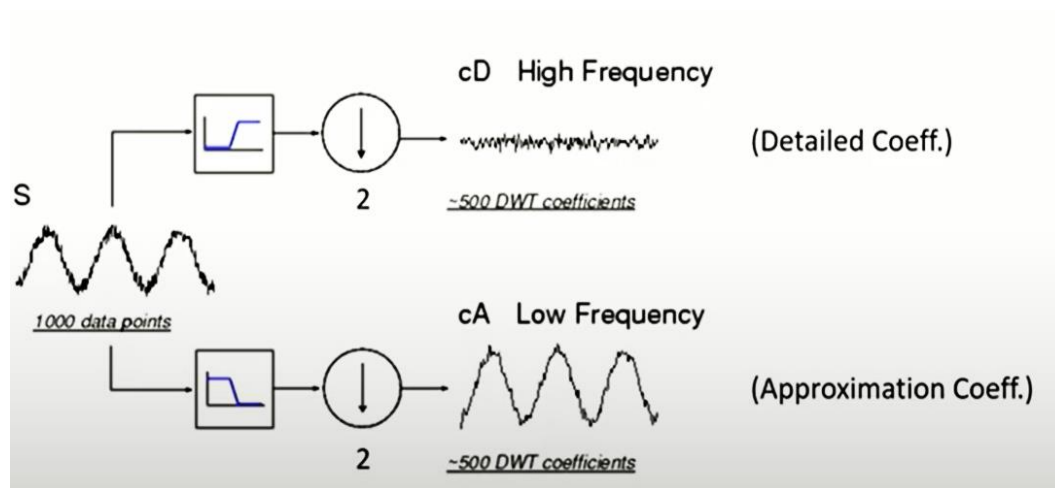
\* Narrow wavelet is translated by small steps.

\* Wide wavelet is translated by large steps.

## ➤ Discrete Wavelet Transform

$$DWT(m, n) = f < f, \psi_{m,n} > = a_0^{-\frac{m}{2}} \sum_{k=-\infty}^{\infty} f(k) \cdot \psi^*(a_0^{-m}(k - nb_0))$$

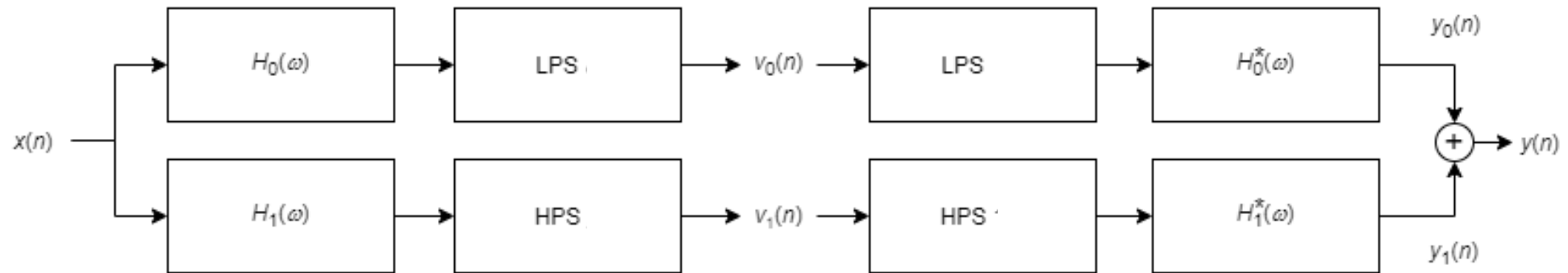
# Wavelet Transform: Filter Bank





# Tunable Q Wavelet Transform Filter Bank

TQWT Analysis and Synthesis Filter Banks



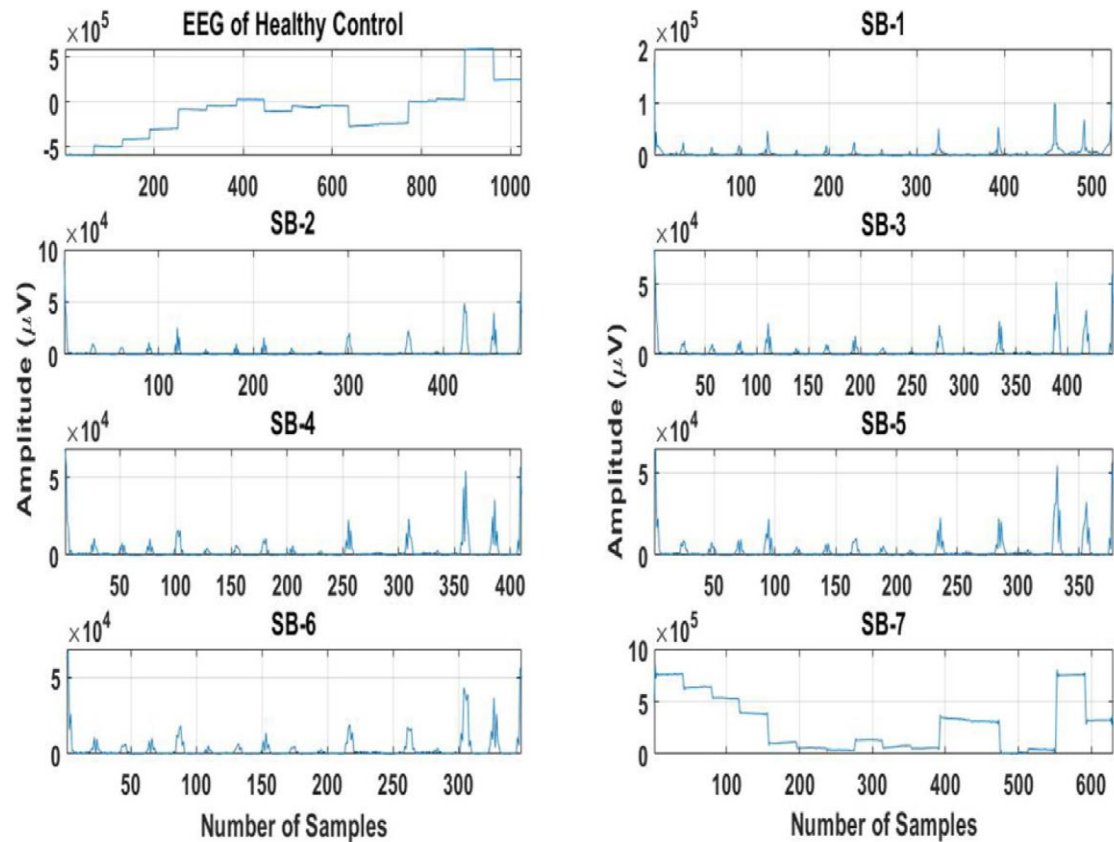
Redundancy Rate:  $\rho = \frac{\xi}{1 - \psi}$

Quality Factor:  $Q = \frac{2 - \xi}{\xi}$

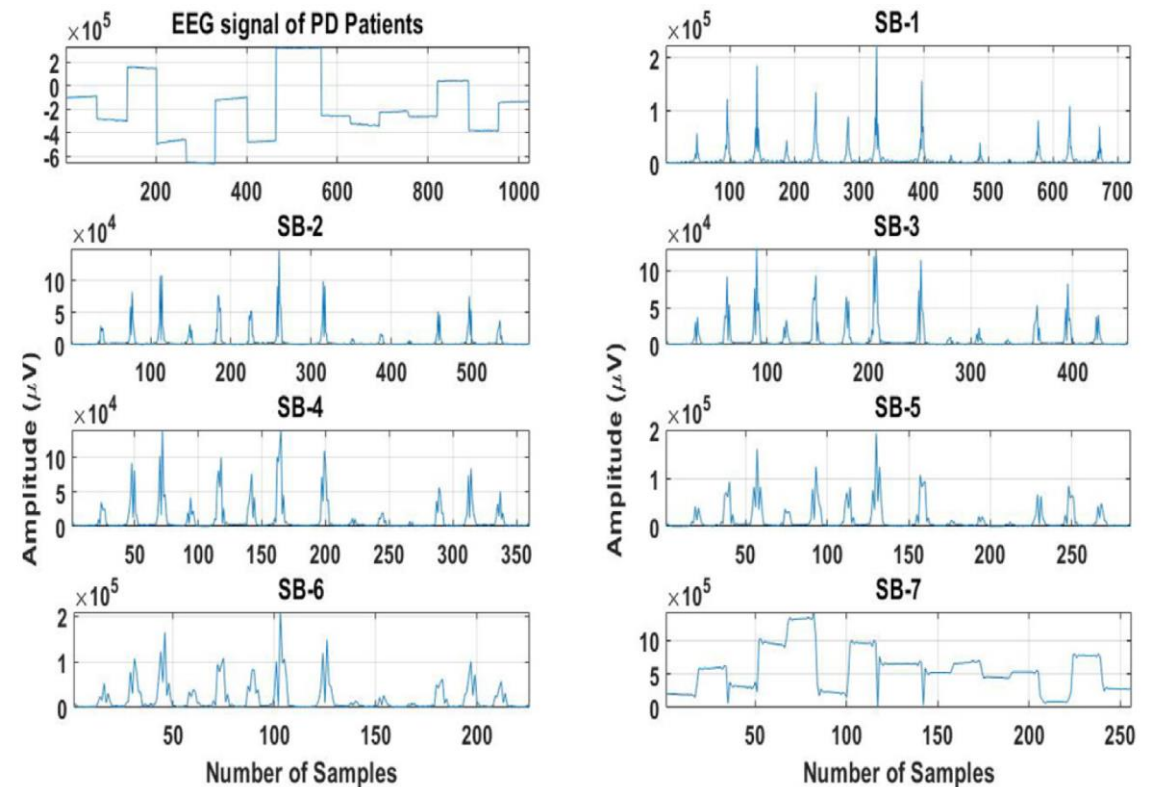
Maximum Number of Levels:  $R_{max} = \left\lceil \frac{\log\left(\frac{N}{4(Q+1)}\right)}{\log\left(\frac{Q+1}{Q+1-\frac{2}{\rho}}\right)} \right\rceil$

\* LPS: Low Pass Scaling ( $\psi$ ), HPS: High Pass Scaling ( $\xi$ ), Number of Sample (N)

# A-TQWT Signal Decompositions

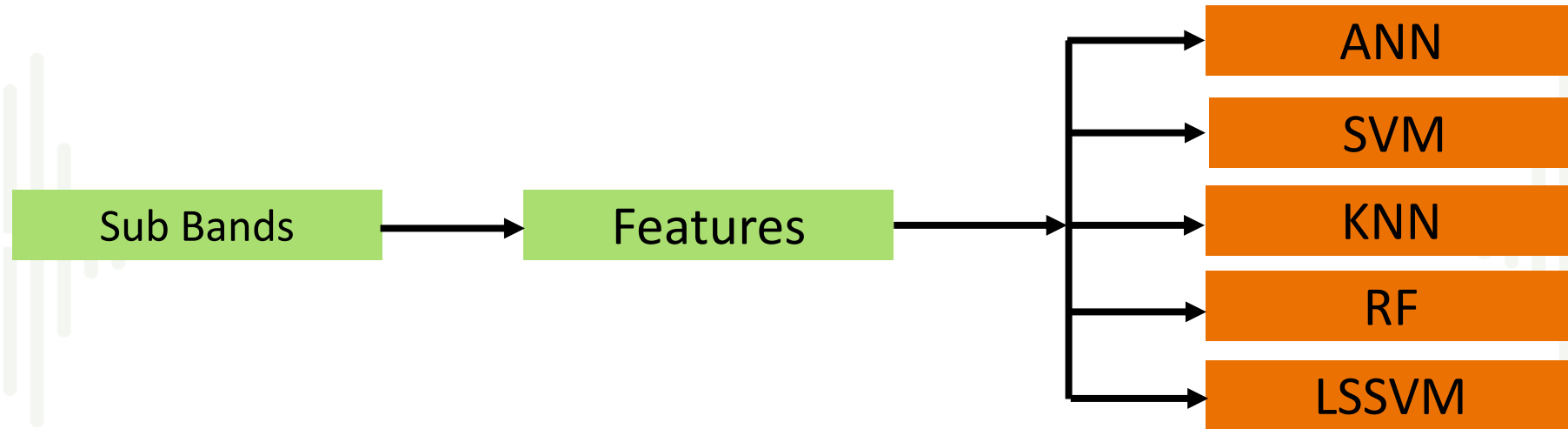


Typical EEG signals: (a) HC subject, and (b) its ATQWT sub-bands.



Typical EEG signals (a) PD subject, and (b) its ATQWT sub-bands

# Classification



**Table-1:** Comparison of average errors obtained due to TQWT and A-TQWT techniques.

Class	TQWT	A-TQWT
HC	0.0491	$3.88 \times 10^{-06}$
SF	0.0456	$1.78 \times 10^{-05}$
SO	0.08903	$1.10 \times 10^{-06}$

\* HC: Health Control

\* SF: PD [Off Medication]

\* SO: PD[On Medication]

**Table-2:** Summary of comparison for automated detection of PD with state-of-the-art techniques.

<b>Authors</b>	<b>Dataset</b>	<b>Method</b>	<b>Classifier</b>	<b>ACC(%)</b>
[1]	Speech	WT, MFCC	SVM	86.64
[2]	EEG	WT	LDD	79.16
[3]	EEG	FFT	RNN	74
[4]	EEG	TVD	SVM	94.34
Proposed	EEG	A-TQWT	LSSVM	<b>97.65, 96.13</b>

# Conclusion

- The proposed A-TQWT provides representative decomposition and reconstruction for signal analysis and synthesis.
- The proposed combination of A-TQWT and LSSVM is so effective that it outperforms existing speech, motions, and EEG-based state-of-the-art techniques.
- In future, the proposed method can be used to detect other neurological disorders like autism, Alzheimer's disease, attention deficit hyperactivity disorder and insomnia.

# References

- 1] K. Smith, B. Varun, A. U. Rajendra, "Detection of Parkinson's disease using automated tunable Q wavelet transform technique with EEG signals," *Journal of Biocybernetics and Biomedical Engineering*, vol. 41, pp. 679-689, 2021.
- [2] O. Wei, H. Basah, H. Lee, V. Vijeane, "Empirical Wavelet Transform Based Features for Classification of Parkinson's Disease Severity," *Journal of Medical Systems*, vol. 42, no. 12, 2017.
- [3] T. B. Drissi, S. Zayrit, B. Nsiri, "Ammoummou A. Diagnosis of Parkinson's Disease Based on Wavelet Transform and Mel Frequency Cepstral Coefficients," *International Journal Adv Computer Science Application*, vol. 10 no. 3, 2019.



Thank You

