



Wavelets

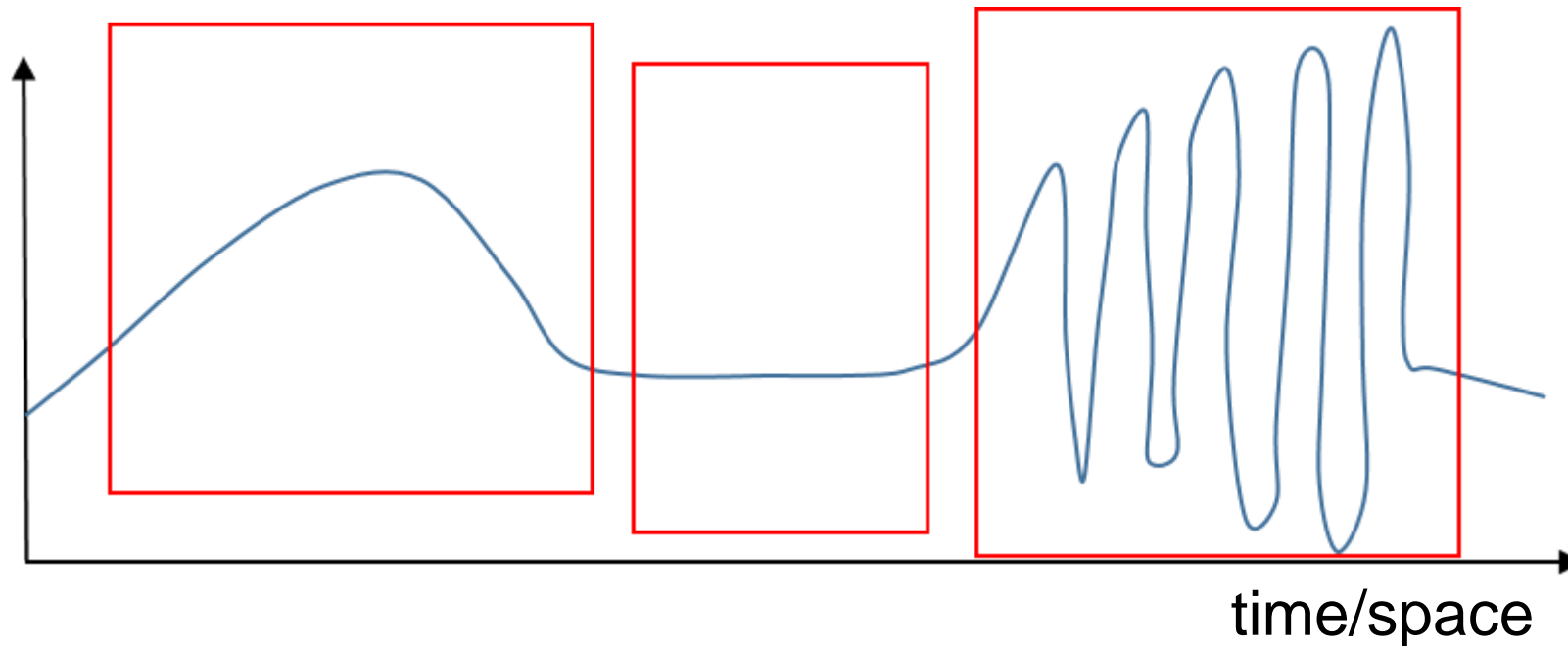
INTELLIGENT SYSTEMS FOR PATTERN RECOGNITION (ISPR)

DAVIDE BACCIU – DIPARTIMENTO DI INFORMATICA - UNIVERSITA' DI PISA

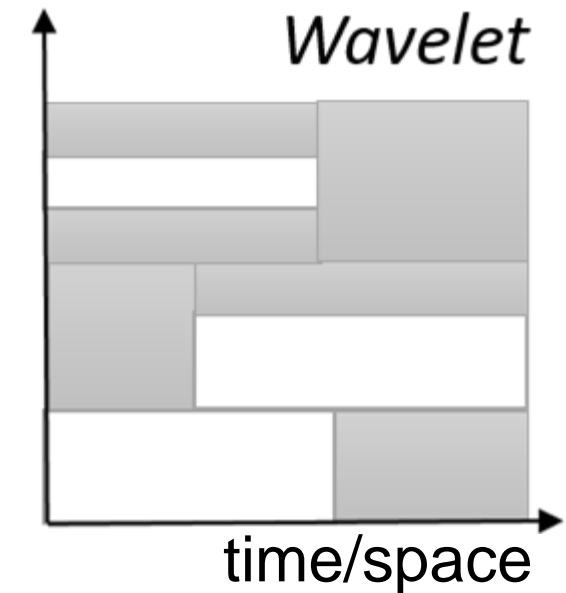
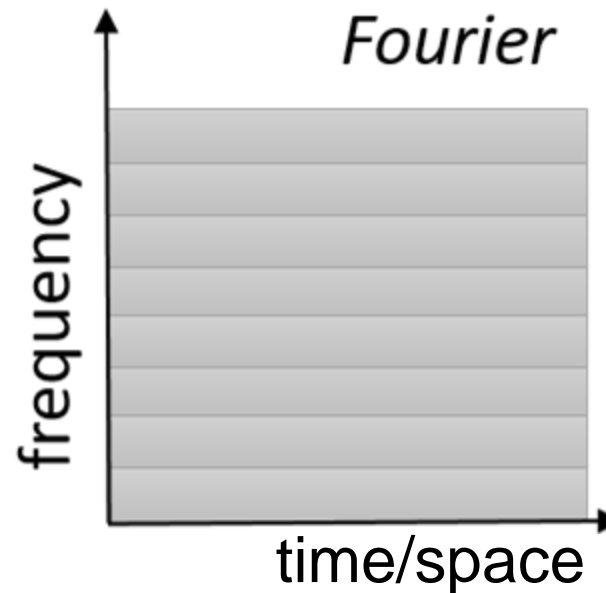
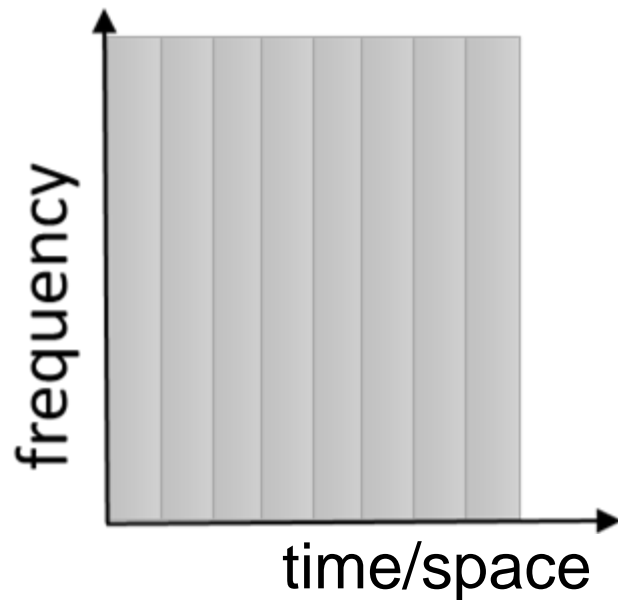
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Limitations of DFT

Sometimes we might need localized frequencies rather than **global frequency analysis**

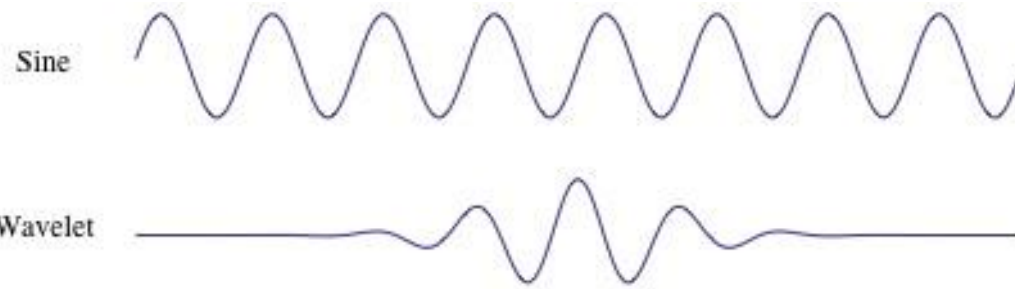


Graphical Intuition



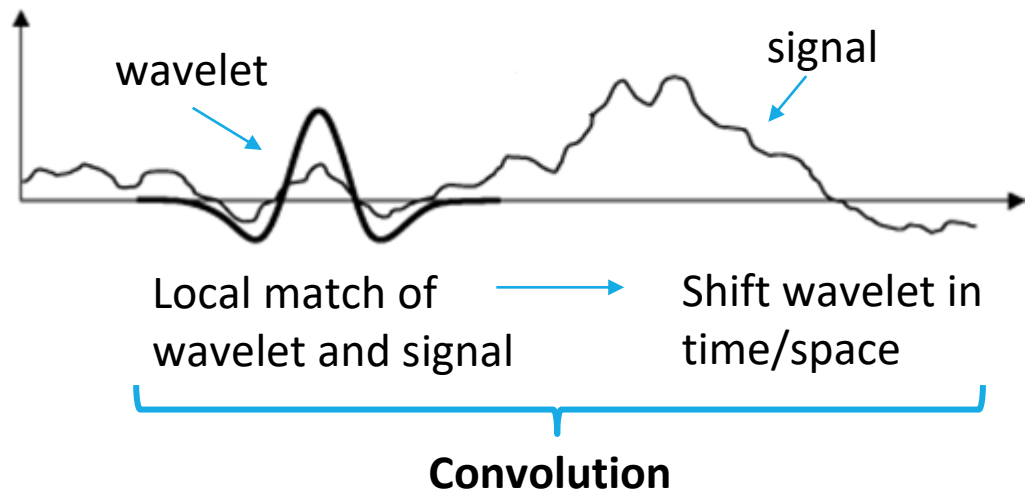
Split signal in frequency bands only if they exist in specific time-intervals or portion of the space

How Does it Work?



Basis function upon which to decompose the signal in **Fourier transform**

Basis function upon which to decompose the signal in **wavelet transform**



1. Scale and shift original signal
2. Compare signal to a wavelet
3. Compute a coefficient of similarity



Wavelets

Split the signal using an orthonormal basis generated by **translation and dilation** of a **mother wavelet**

$$\sum_t x(t) \Psi_{j,k}(t)$$

Terms k and j regulate scaling and shifting of the wavelet

$$\Psi_{t,k}(t) = \frac{1}{\sqrt{2^k}} \Psi((t - j)/2^k)$$

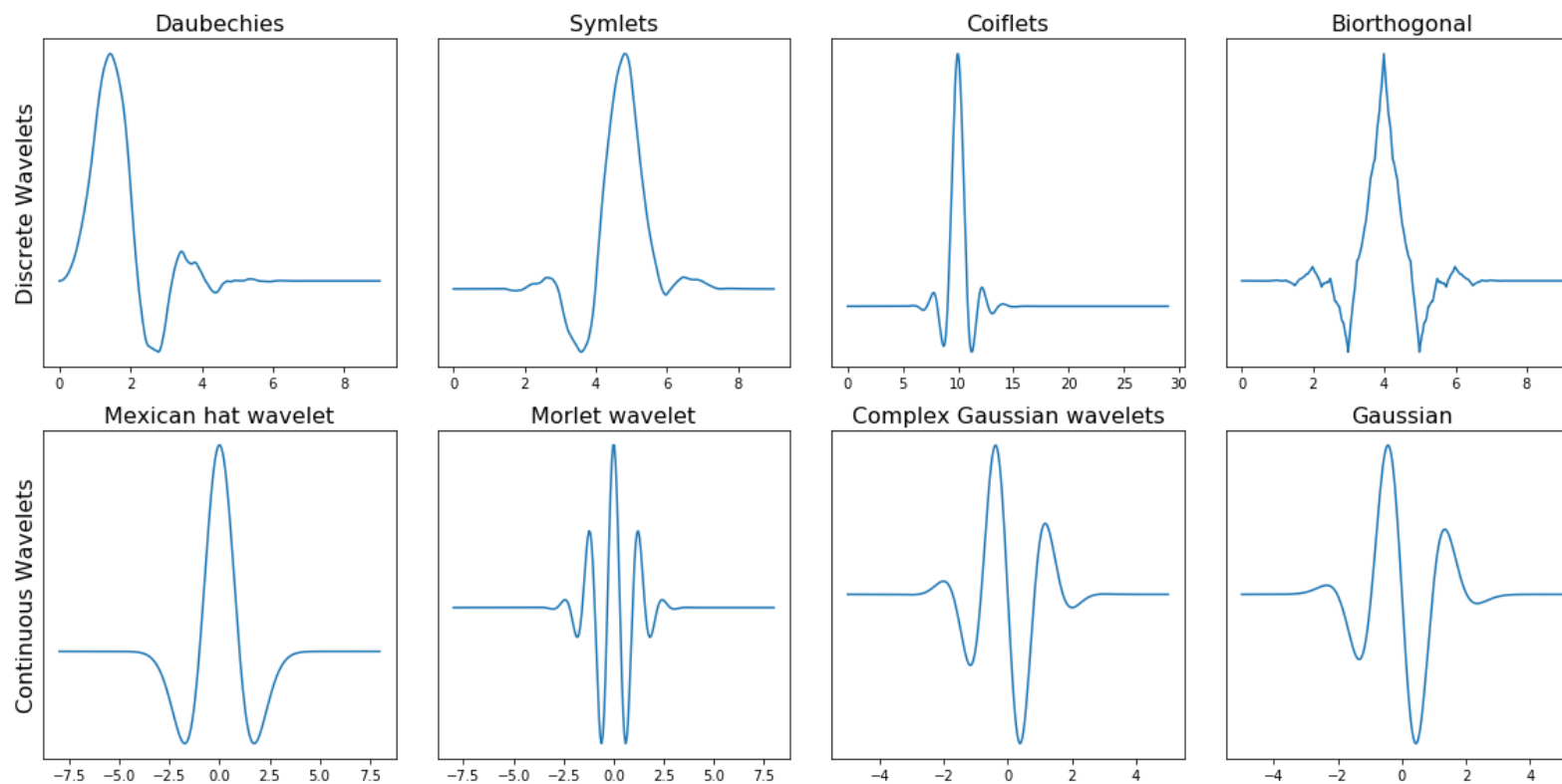
with respect to the mother $\Psi(\cdot)$.

- $k < 1$ Compresses the signal
- $k > 1$ Dilates the signal

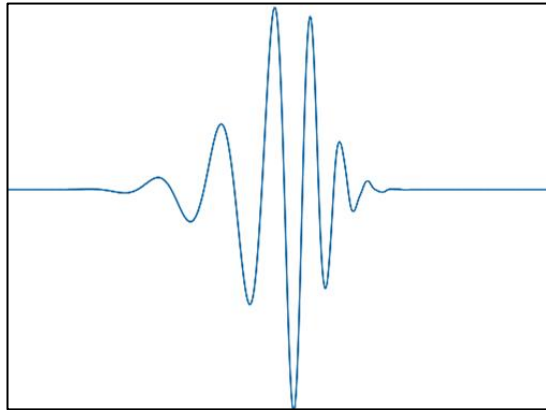


A (partial) wavelet dictionary

Many different possible choices for the mother wavelet function

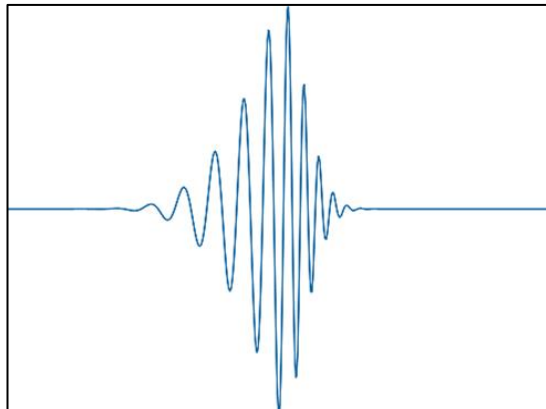


Scaling/dilation is akin to (sort of) frequency



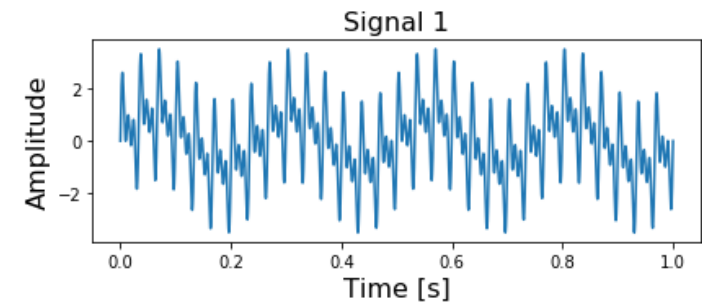
High scale

- Stretched wavelet
- Slowly changing, coarse features
- Low frequency

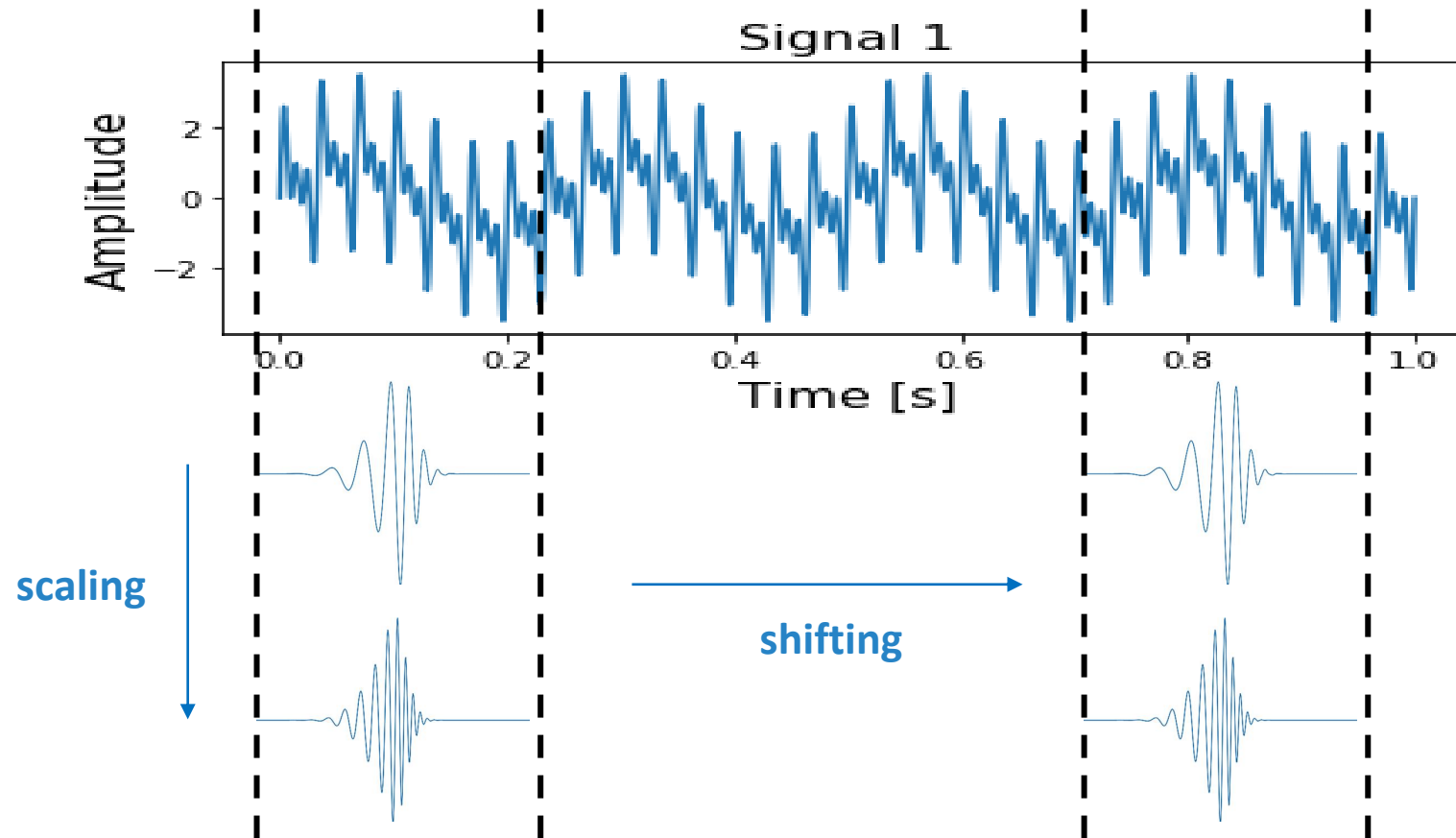


Low scale

- Compressed wavelet
- Rapidly changing details
- High frequency

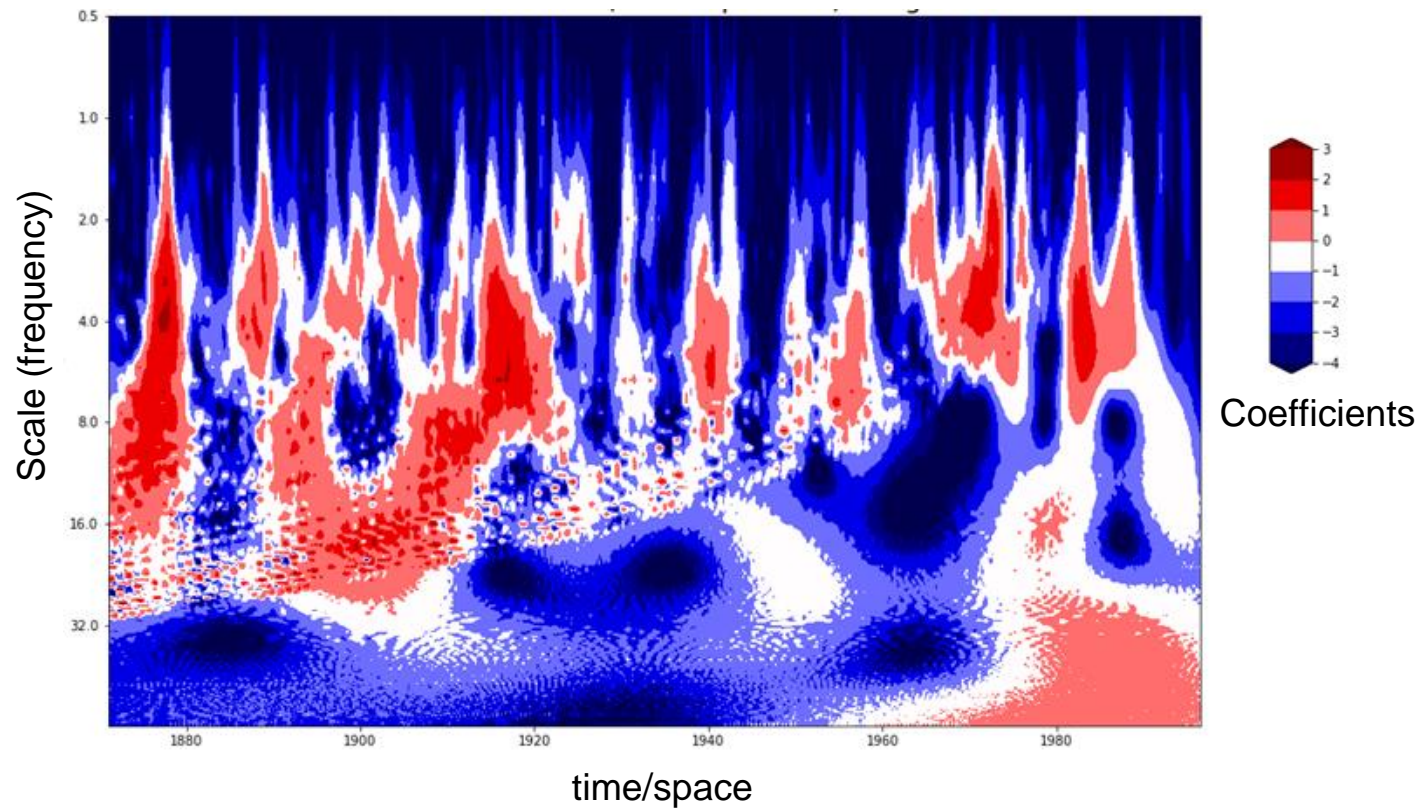


Shifting moves the wavelet in time/space on the signal



Compute coefficients of signal-wavelet match across all scales and shifts

Coefficient Plot



This is once-again the **power-spectrum** of the signal, revealed by the (continuous) **Wavelet Transform**

Discrete Wavelet Transform (DWT)

- Use a **finite set of scales and shifts** rather than “any possible value” as in the continuous wavelet transform
- Subset scale continuous values **using power-of-two values with step 1** (and translates proportionally to scale if **decimated**)
- **Key aspects**
 - Efficient and sparse representation
 - Orthonormal basis
 - Can always be inverted



Using the WT in PR Applications (I)

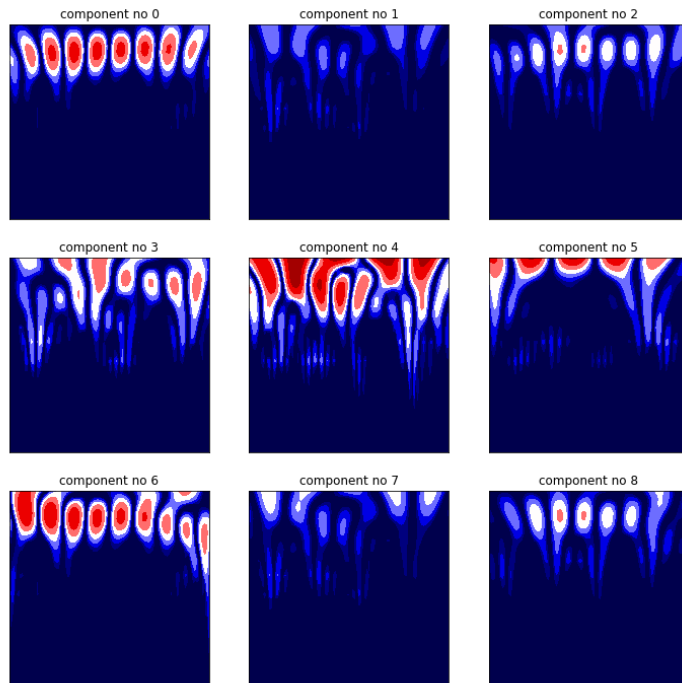
Human Activity Recognition Using Smartphones Dataset
(Reyes-Ortiz et al, 2012)



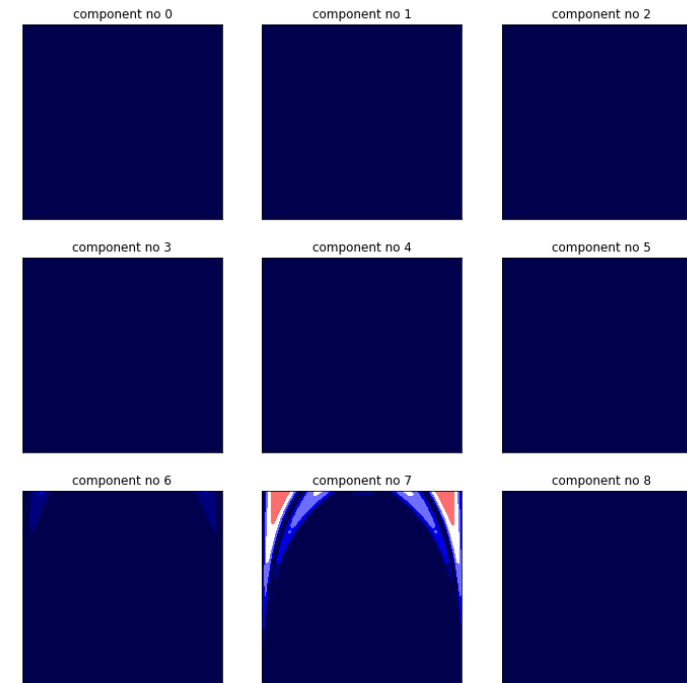
9-components sensor measurements of people doing different activities (walking, laying, standing, ...)

Using the WT in PR Applications (II)

Activity: walking upstairs

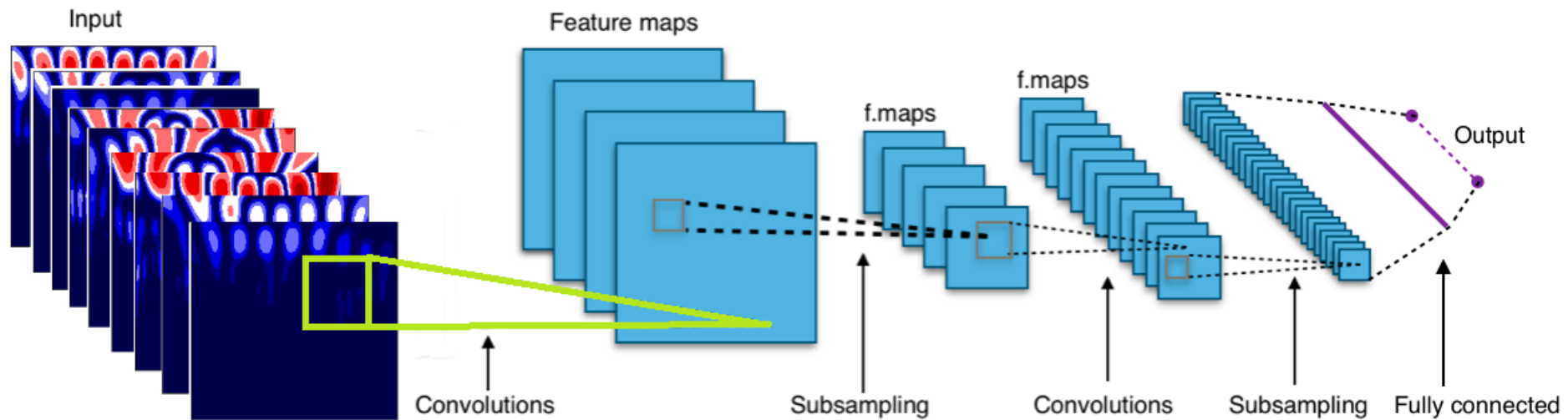


Activity: laying



Spectrograms for different activities alone lead to a classification accuracy of 0.91

Using the WT in PR Applications (III)



Using wavelet transform + convolutional neural networks \Rightarrow 0.96

Code

- [PyWavelets](#) - Wavelet transforms in Python
- [Wavelet Toolbox](#) – Wavelet transforms in Matlab

Take Home Messages

- Fourier transform
 - Basis functions: sinusoids
 - Only offers frequency information
- Wavelet transforms
 - Basis functions: small waves (wavelets)
 - Frequency and temporal/spatial information
- Wavelets can be more effective on discontinuous and bursty data



Next Lecture

Generative and Graphical Models

- Introduction to a module of 10 lectures
- A refresher on probabilities
 - Probability theory
 - Conditional independence
 - Inference and learning in generative models
- Graphical models representation
- Directed, undirected and dynamic graphical models

