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OF TECHNOLOGY

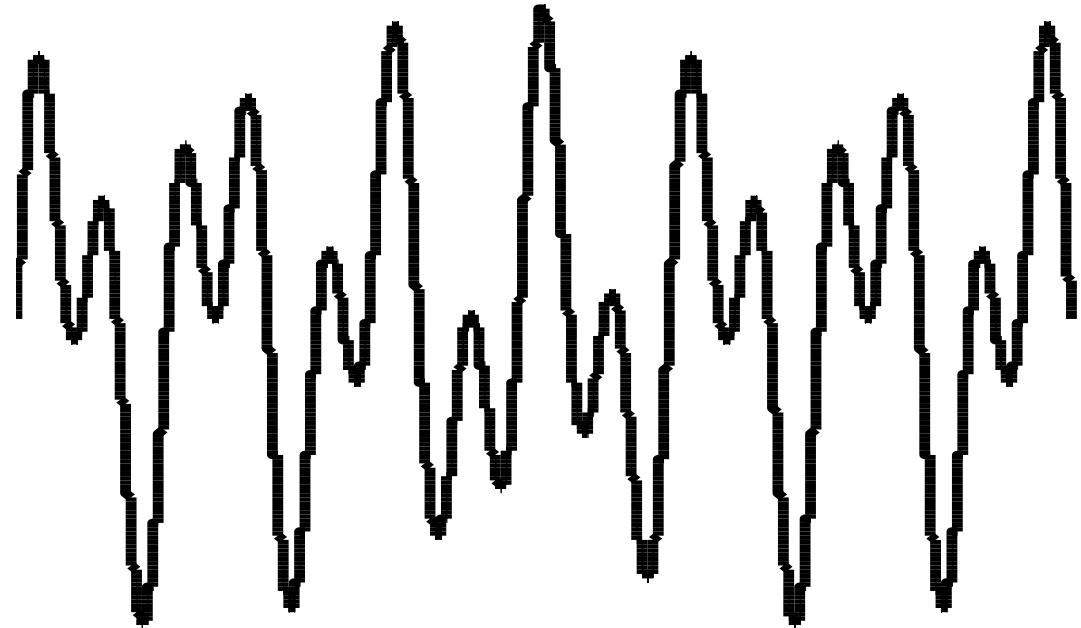
Video Computing

Prof. Haibo Li

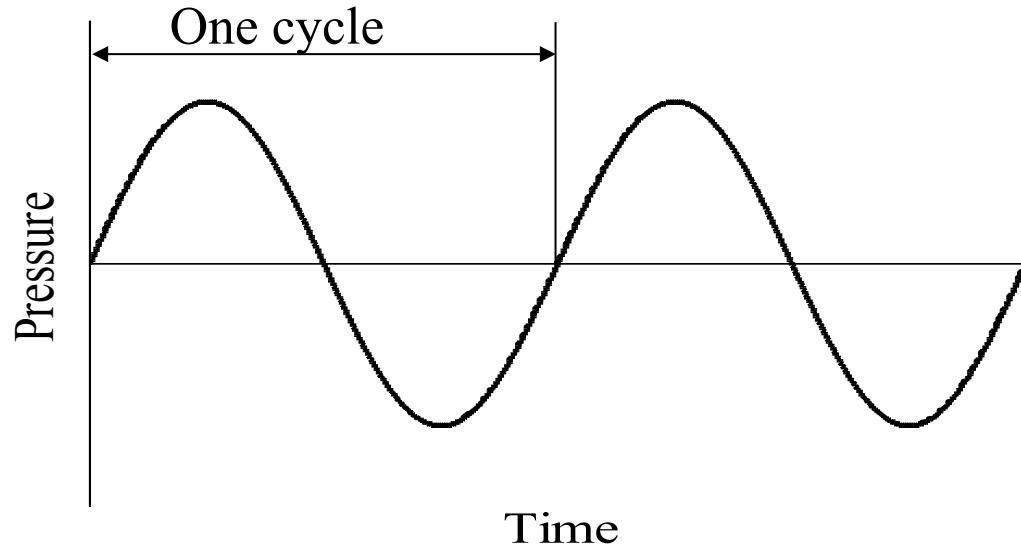


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How to Characterize a Signal?



A Typic Signal



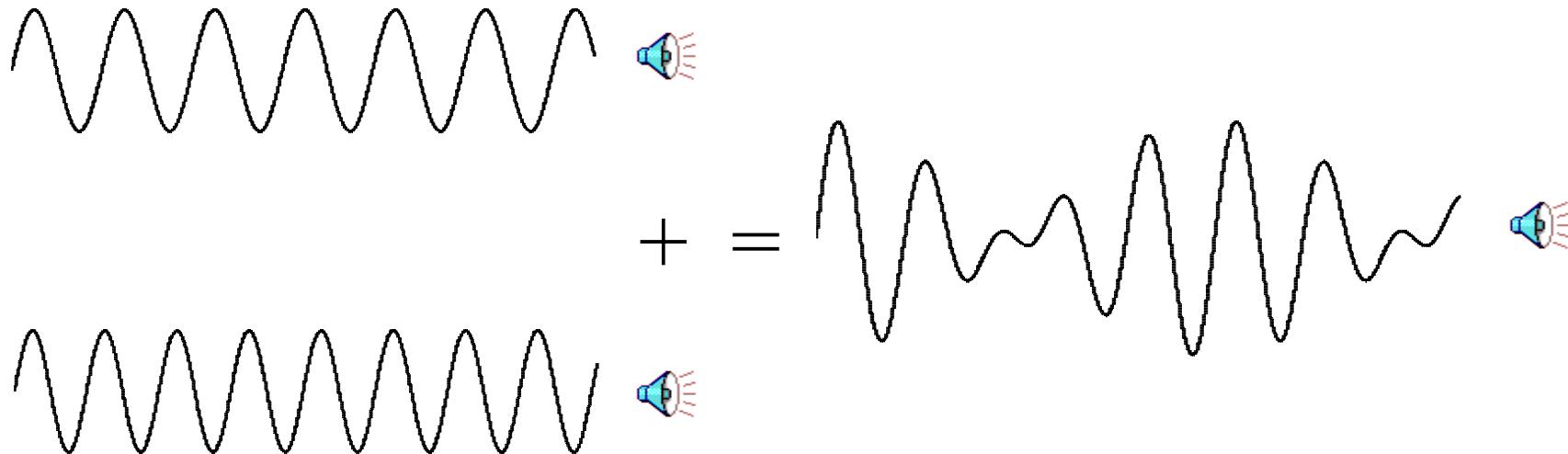
Frequency = number of cycles per second

unit of frequency: Hertz (Hz)

e.g. 400 Hz = 400 cycles per second

Signal Synthesis

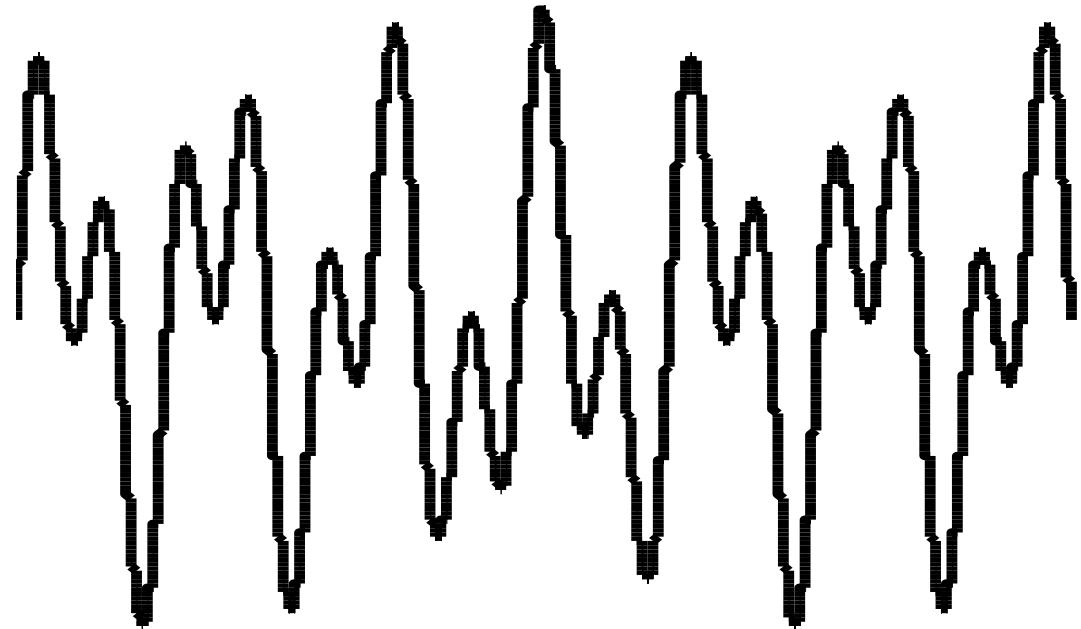
Consider the situation in which we have two tones of different frequencies presented simultaneously.





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How to Characterize a Signal?





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Fourier told us.....

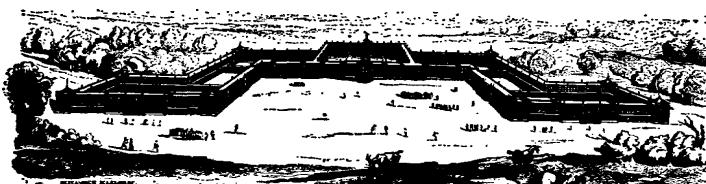


JOURNAL DE L'ÉCOLE SOCIÉTAIRE
TOMBE ENTE AU PHALANGE
LE PHALANSTÈRE OU LA RÉFORME INDUSTRIELLE (1832-1834).

LA
PHALANGE
JOURNAL DE LA SCIENCE SOCIALE
DÉCOUVERTE ET CONSTITUÉE
PAR
CHARLES FOURIER.

Industrie, Politique, Sciences, Art et Littérature.

TOME I.

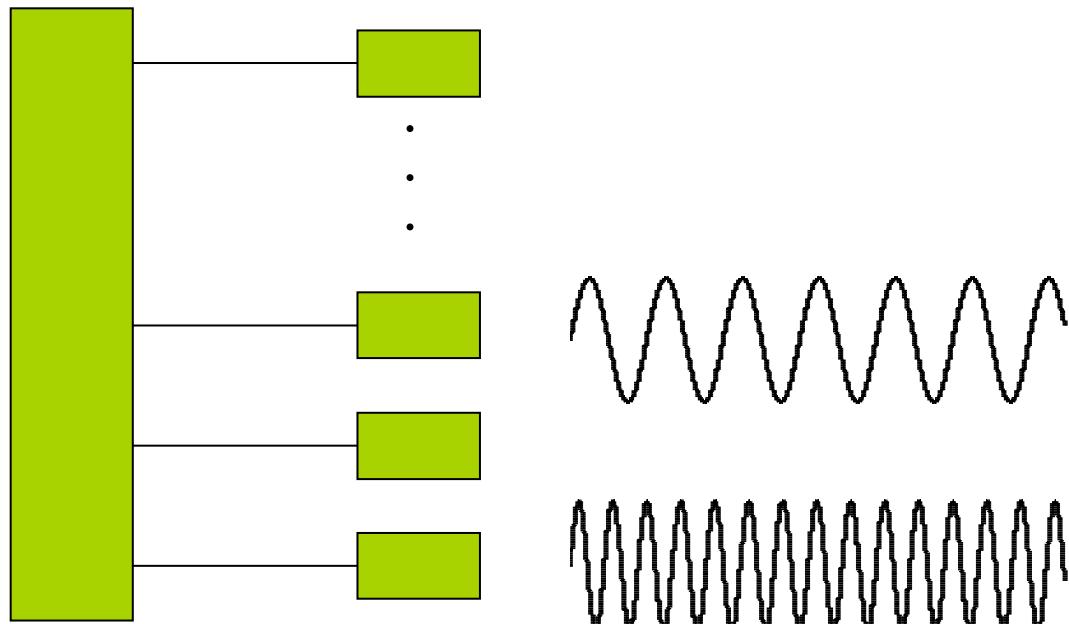
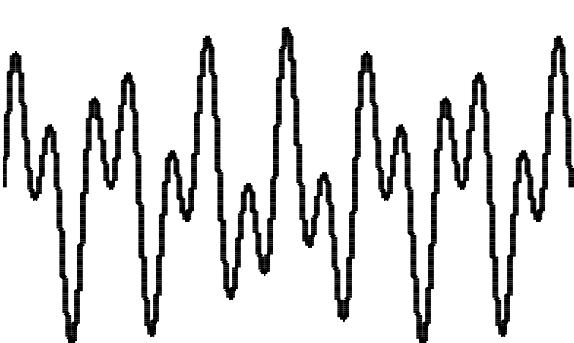


IDEË D'UN PHALANSTÈRE.
Habitation d'une PHALANGE de 400 à 500 familles occupée en fonctions de:
Agriculture, métiers, fabriques, commerce, art, sciences, etc.
Emploiant, dans l'ordre sociétaire, les 400 à 500 constructions incohérentes,
maisons, masure, grange, étable, etc.,
d'une bourgeoisie de 1800 à 2000 habitants dans l'ordre mondial actuel.

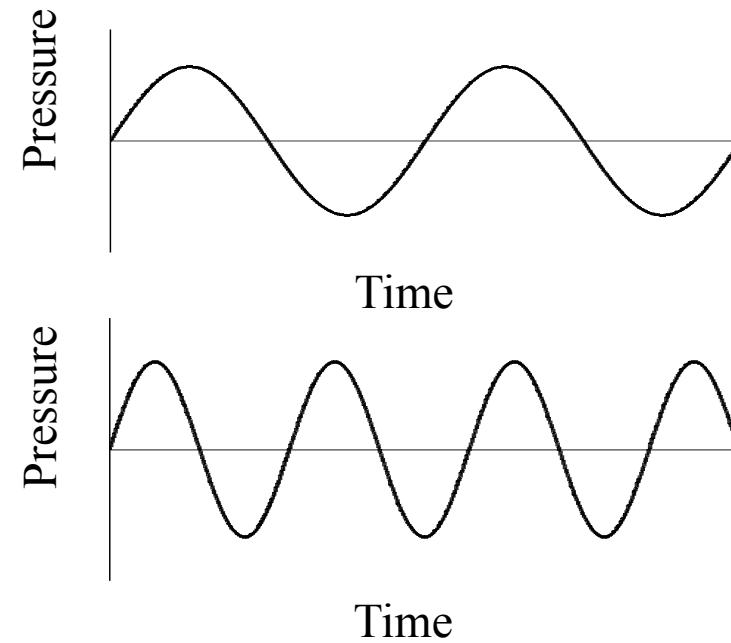
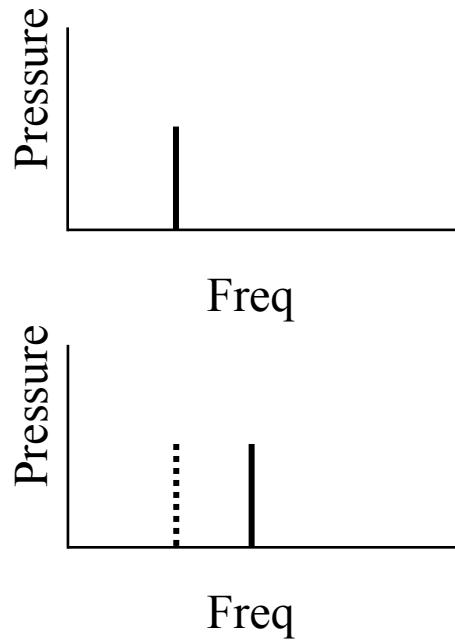
PARIS
AU BUREAU DE LA PHALANGE, RUE JACOB, N° 54.
1836-1837

Signal Analysis

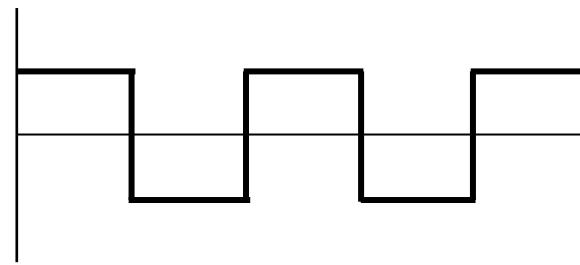
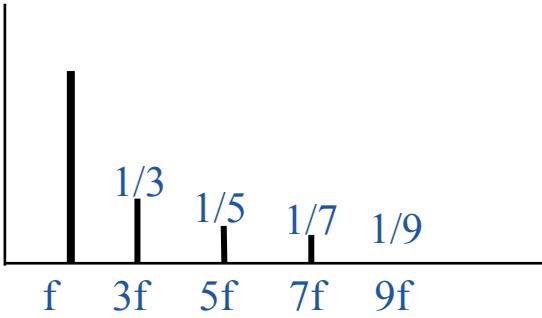
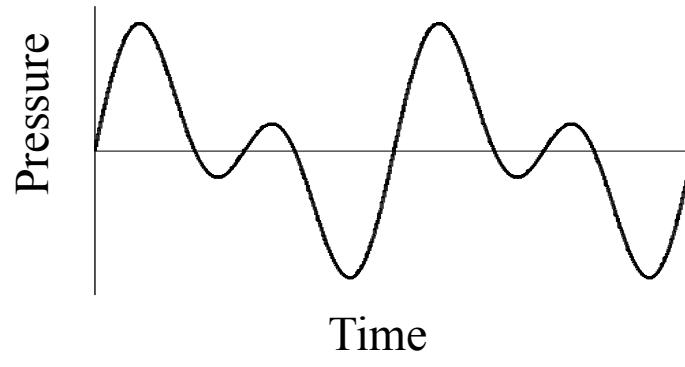
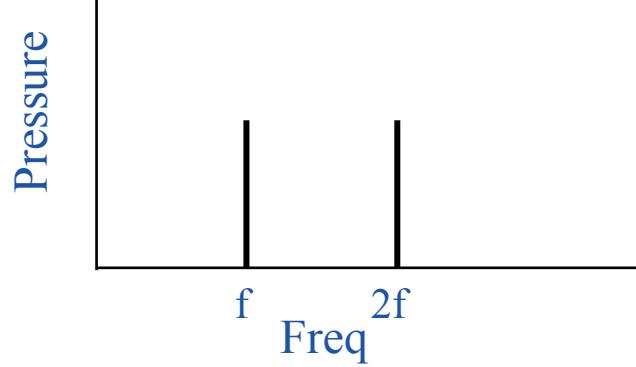
A preliminary analysis of a sound mixture
into individual frequency components



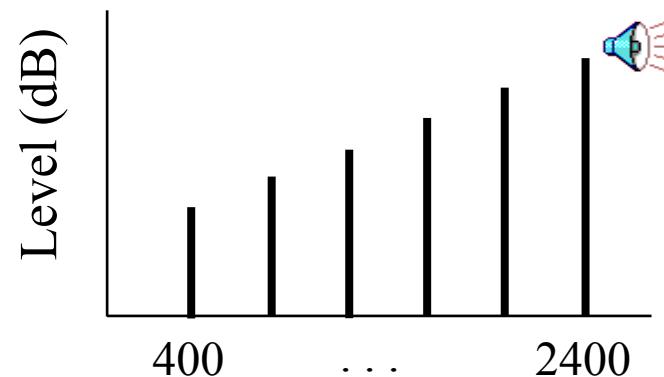
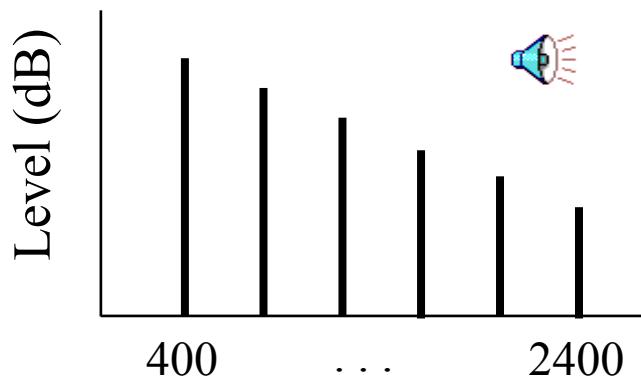
Fourier Transform



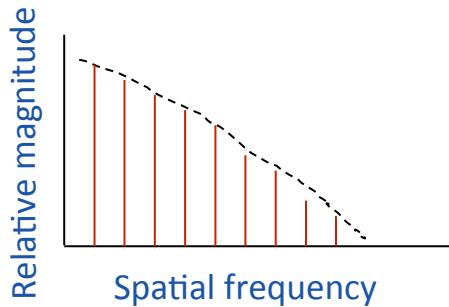
Fourier Transform



Signal Characterization



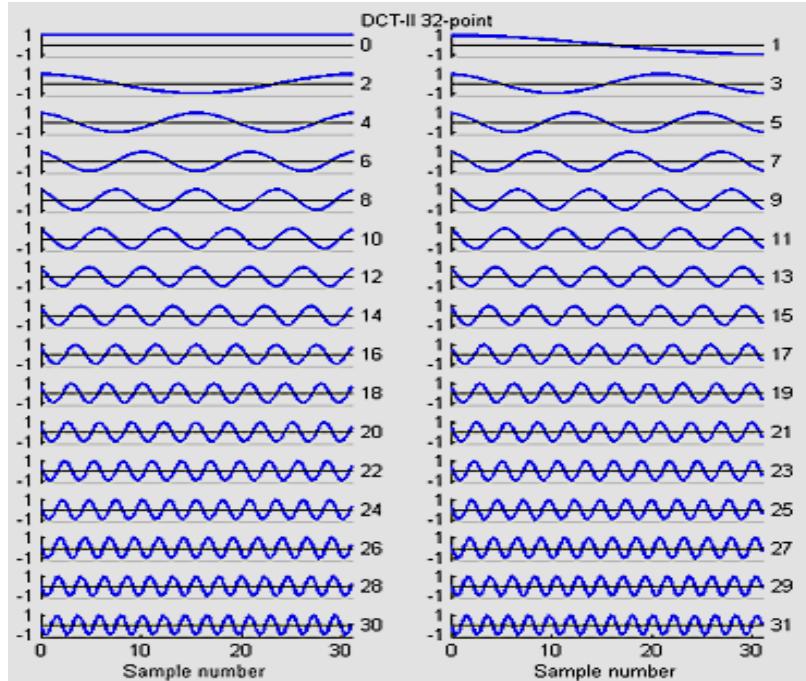
Natural Signals





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Fourier Transform of a Picture

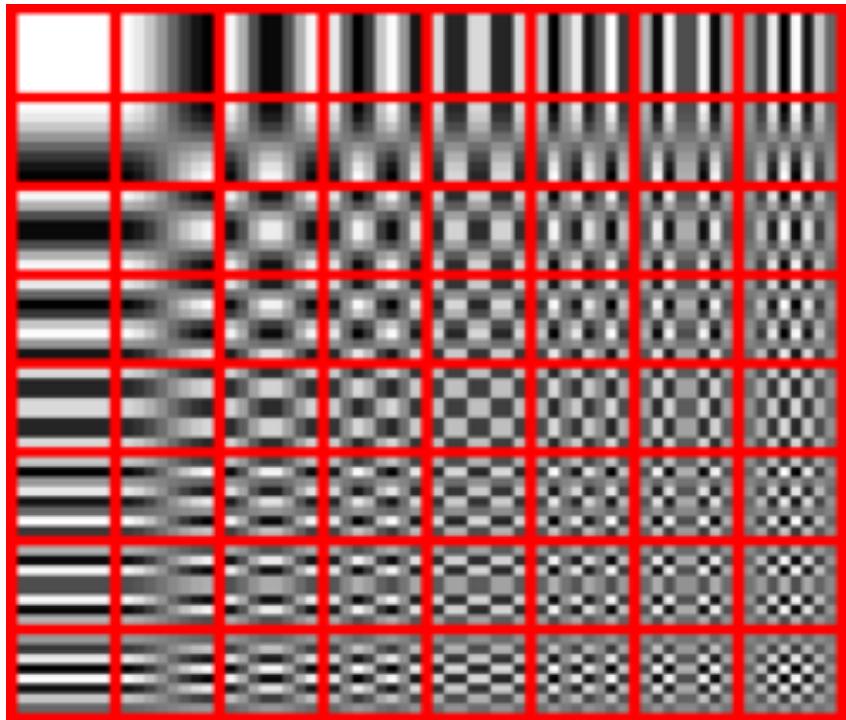


DCT = discrete Cosine Transform (real functions)

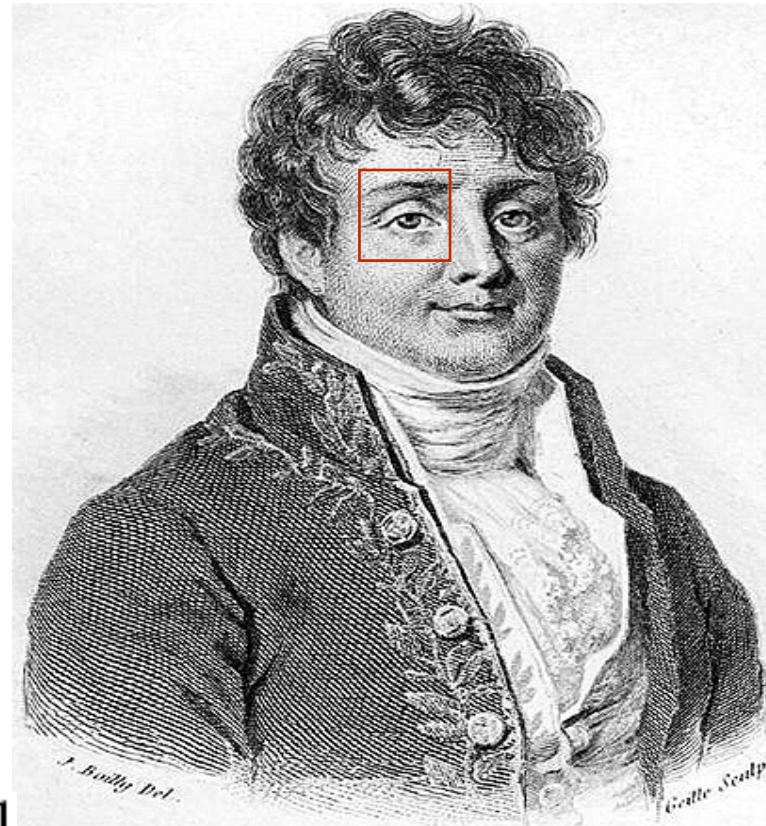


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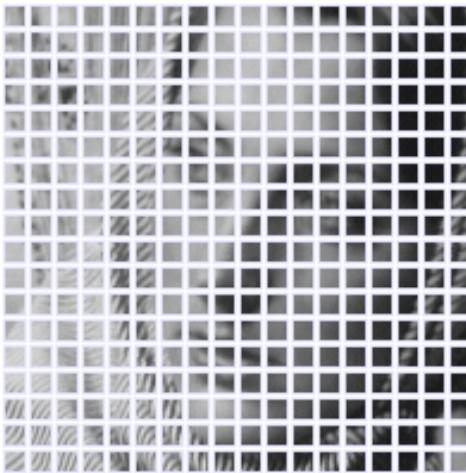
Image Compression: DCT



$$\alpha_p \alpha_q \cos \frac{\pi(2m+1)p}{2M} \cos \frac{\pi(2n+1)q}{2N}, \quad 0 \leq p \leq M-1 \\ 0 \leq q \leq N-1$$

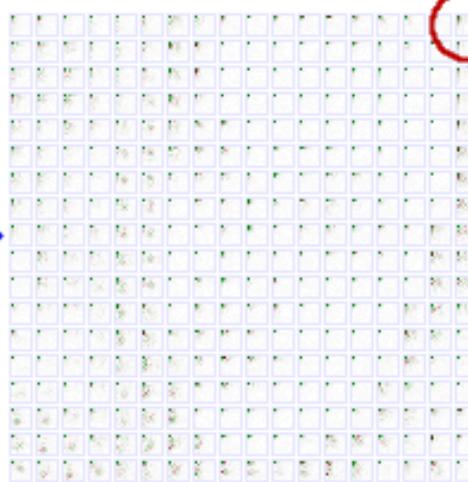


JPEG Image Compression

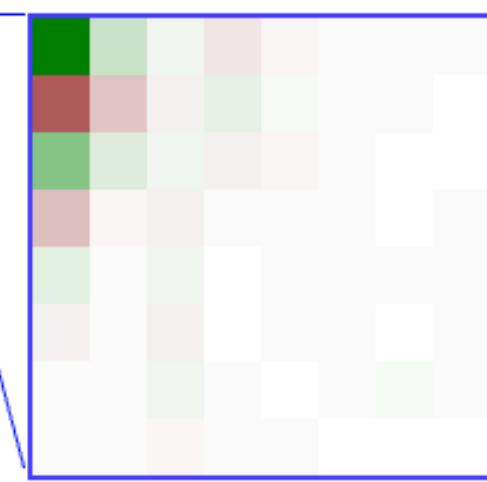


Original image

Pixel blocks

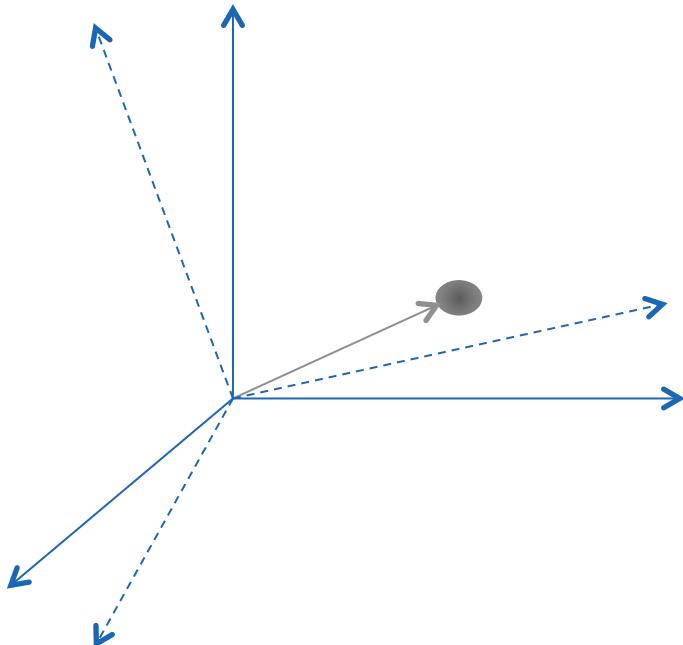


DCT coefficient blocks



Single coefficient block

Signal Representation



$$x = a_0 \mathbf{e}_0 + a_1 \mathbf{e}_1 + a_2 \mathbf{e}_2 + \dots$$

Two questions

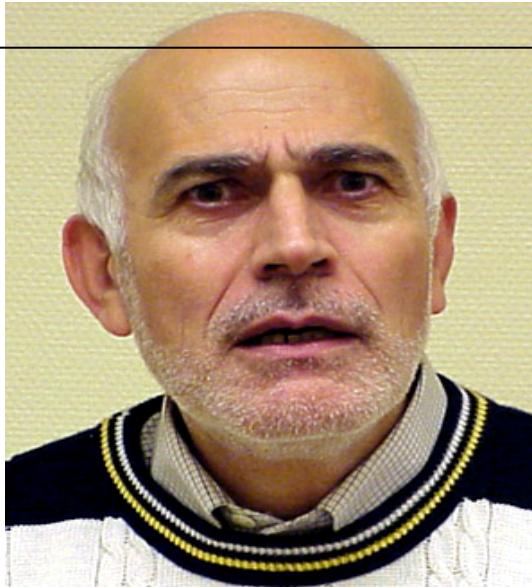
- how to select basis \mathbf{e}_i
- how to calculate a_i



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Vector Representation

123 112 122 124
120 119 122 122
112 130 140 112

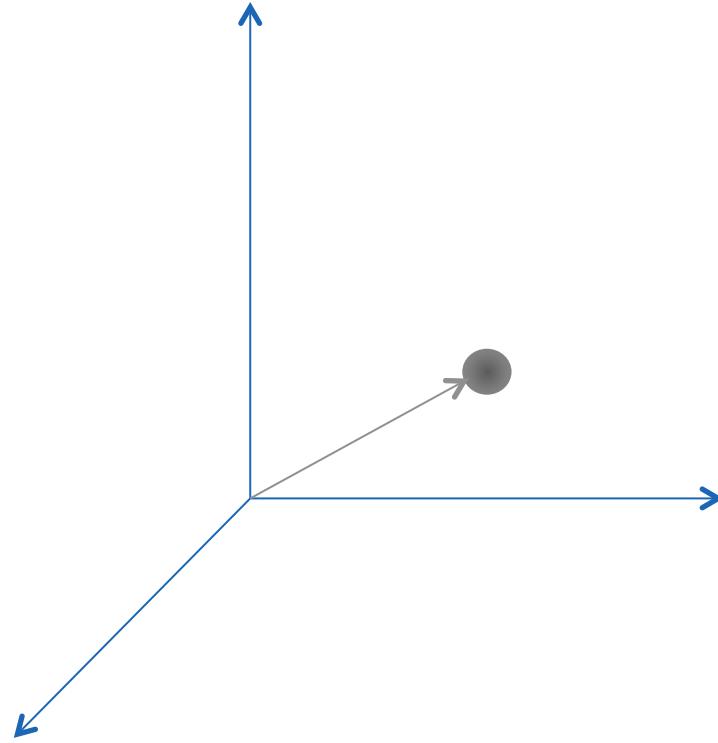




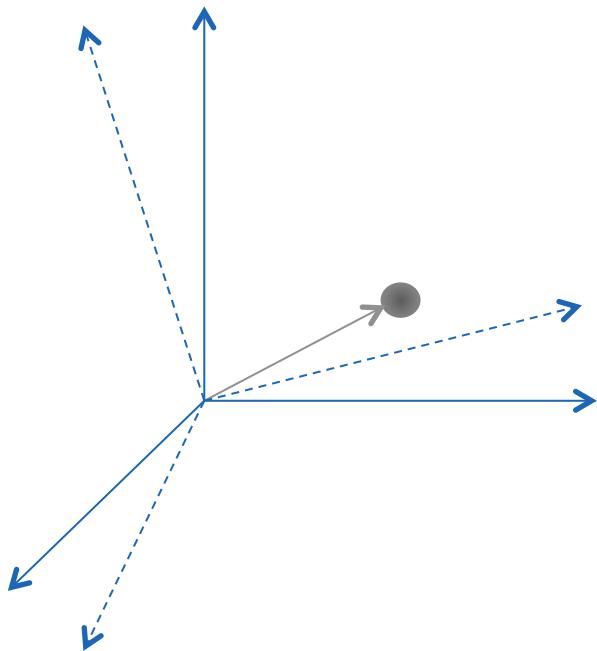
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Media Signals

X =



Signal Representation

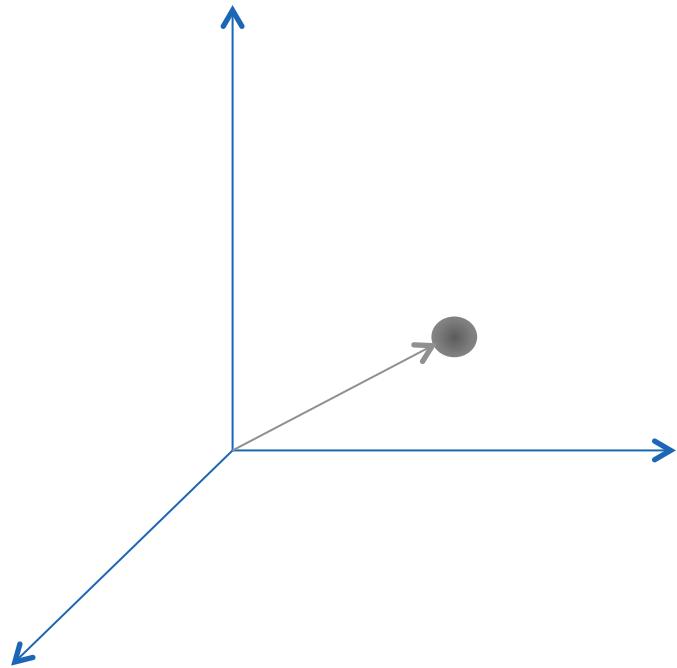




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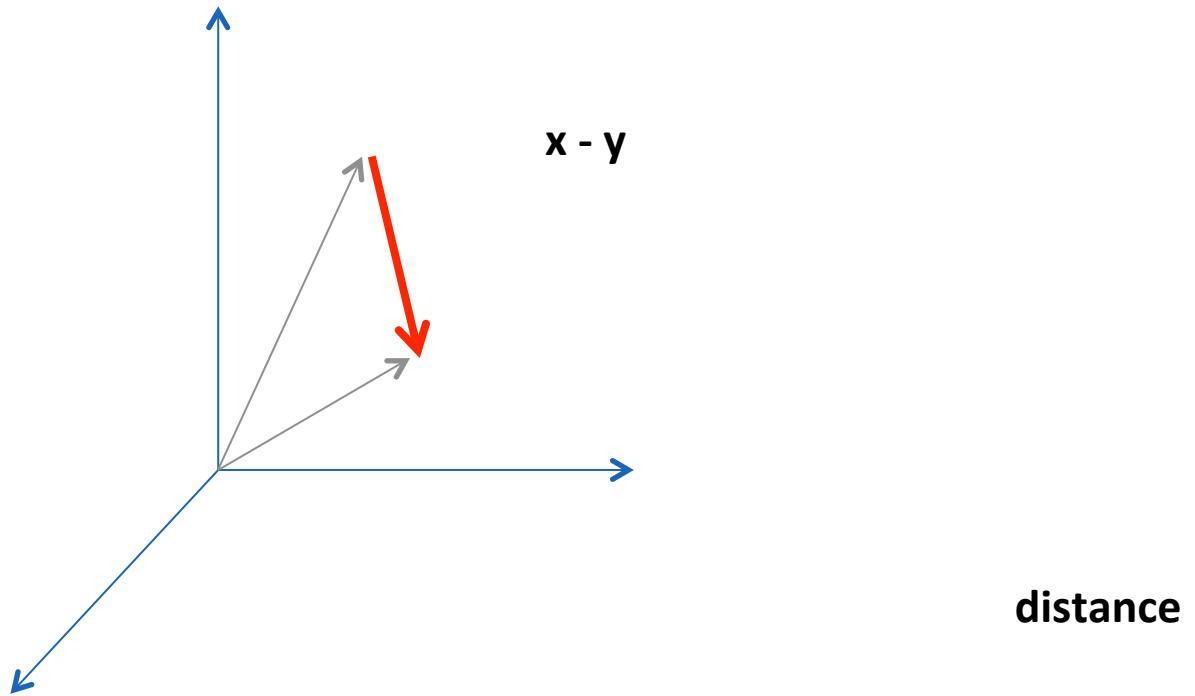
Vector Operations

Signal Representation



magnitude = norm

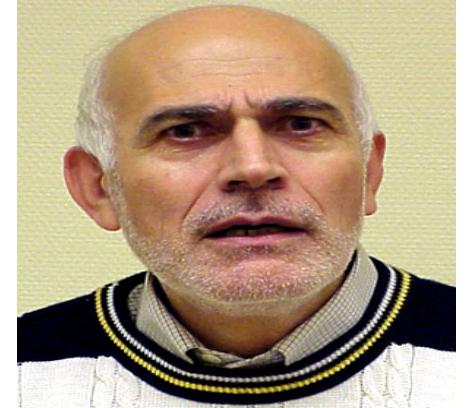
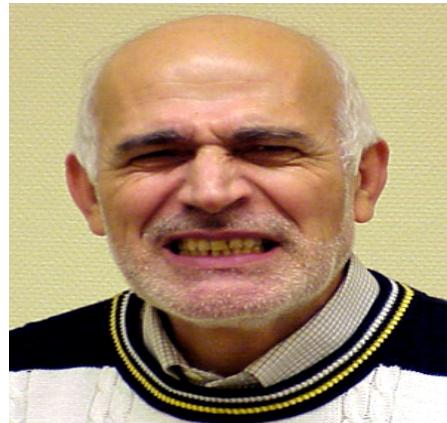
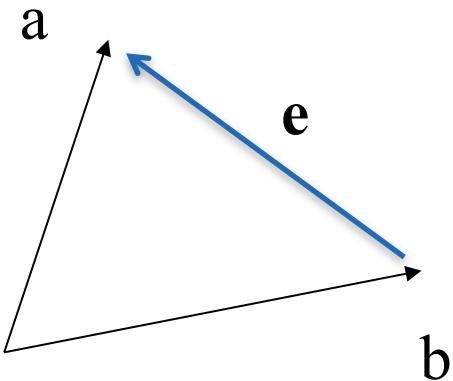
Signal Representation





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Vector Distance



a

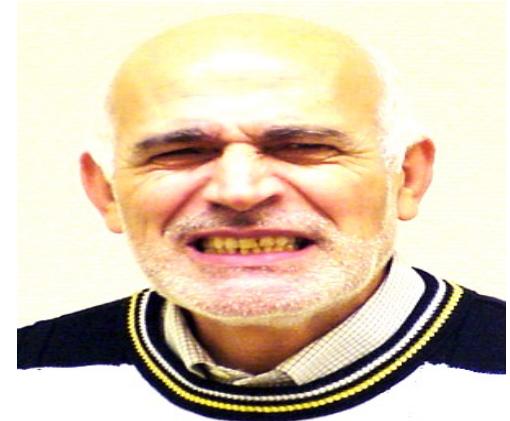
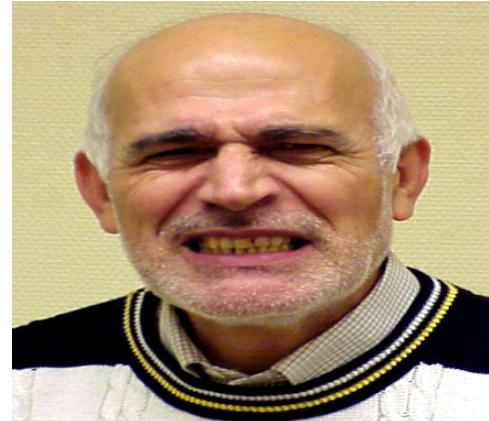
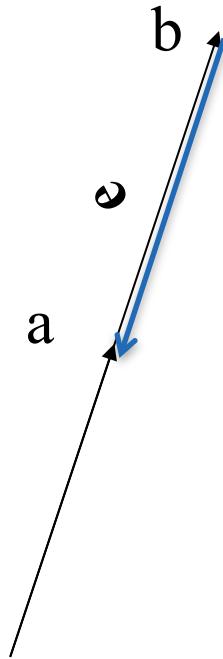
b

$$e = a - b$$



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Vector Distance

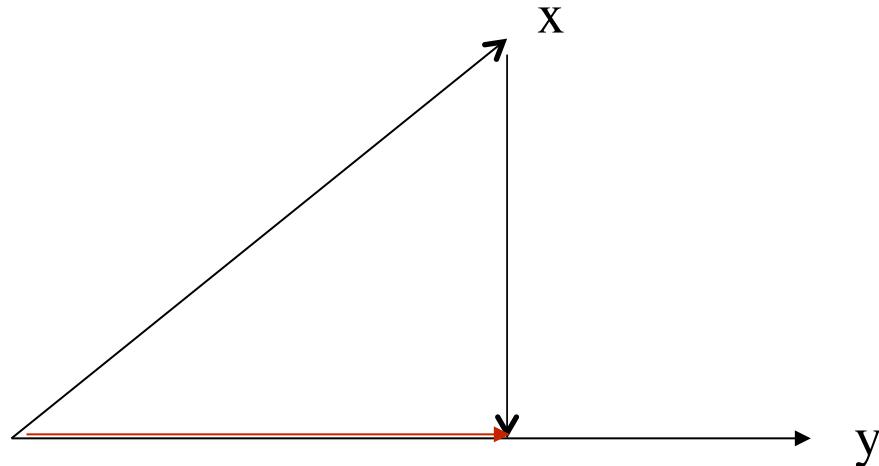


a

b

$$e = a - b$$

Signal Representation



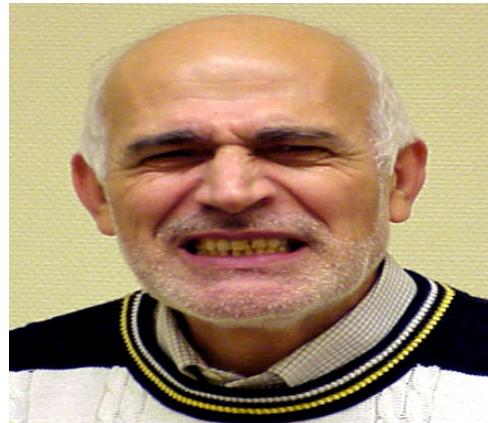
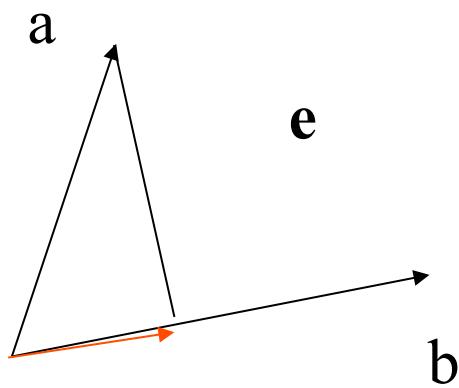
$$\text{projection} = \mathbf{x}\mathbf{y}^T$$



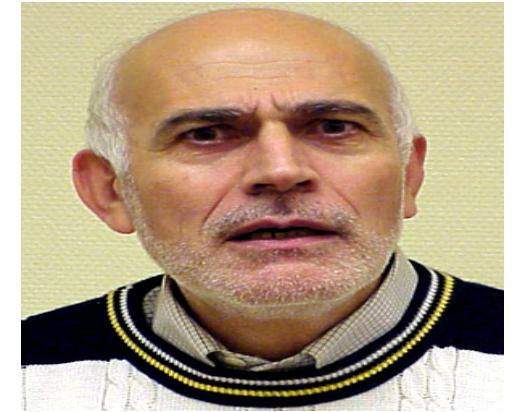
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Vector Approximation

$$e = a - \tilde{a}$$



a



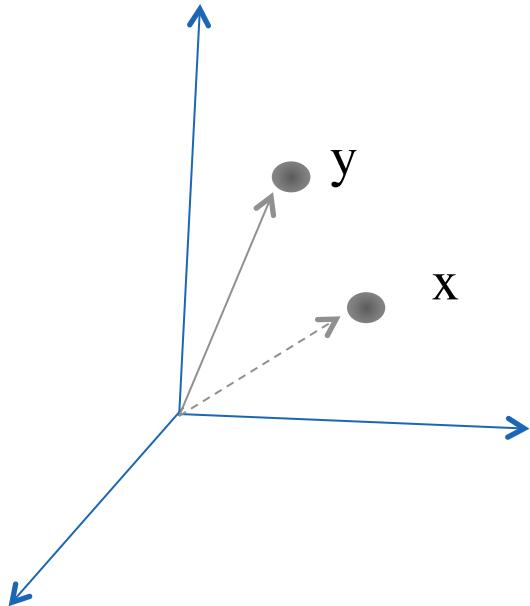
b

$$\tilde{a} = kb$$

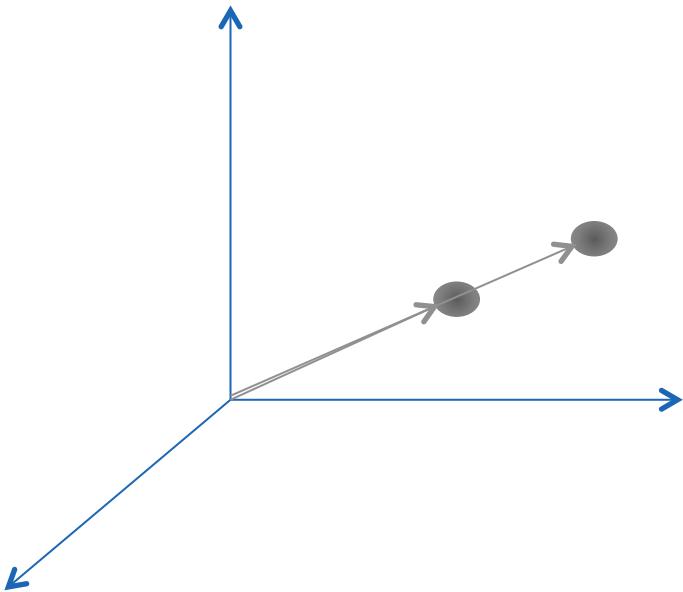
Signal Representation

Linear Transform

$$\mathbf{y} = \mathbf{Ax}$$



Signal Representation



Linear Transform

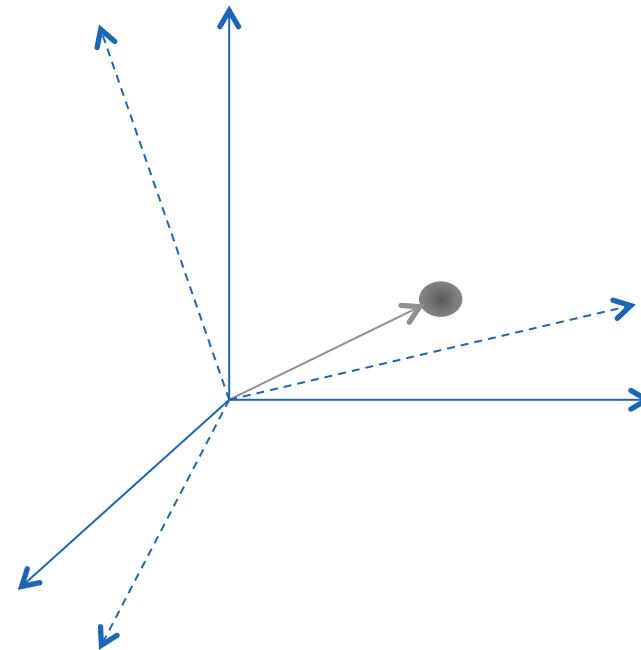
$$\mathbf{y} = \mathbf{Ax} = \lambda \mathbf{x}$$

Eigen value
and
Eigen vector

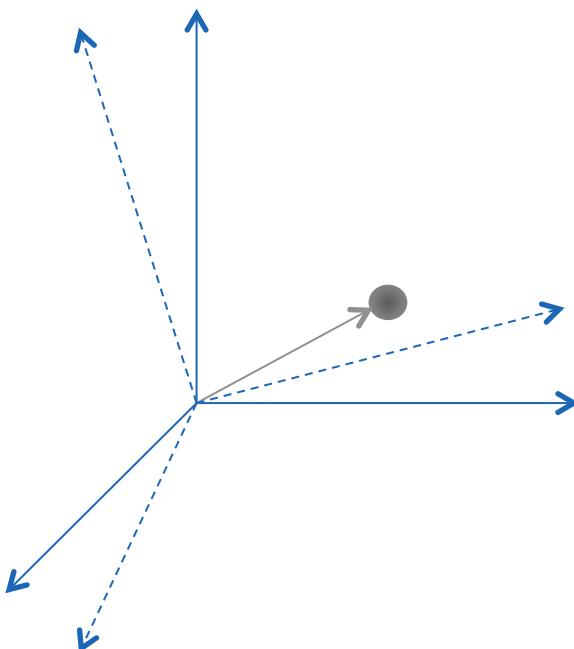


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Signal Representation



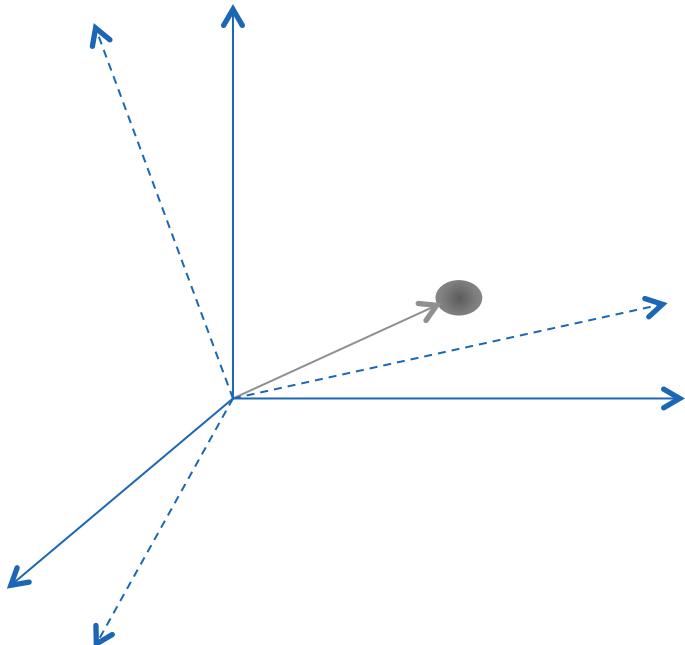
Signal Representation



$$x = a_0 \mathbf{e}_0 + a_1 \mathbf{e}_1 + a_2 \mathbf{e}_2 + \dots$$

$$\begin{matrix} | & = & [& | & | & | &] & | \\ | & & | & | & | & | & | & | \end{matrix}$$

Signal Representation



$$x = a_0 \mathbf{e}_0 + a_1 \mathbf{e}_1 + a_2 \mathbf{e}_2 + \dots$$

Two questions

- how to select basis \mathbf{e}_i
- how to calculate a_i



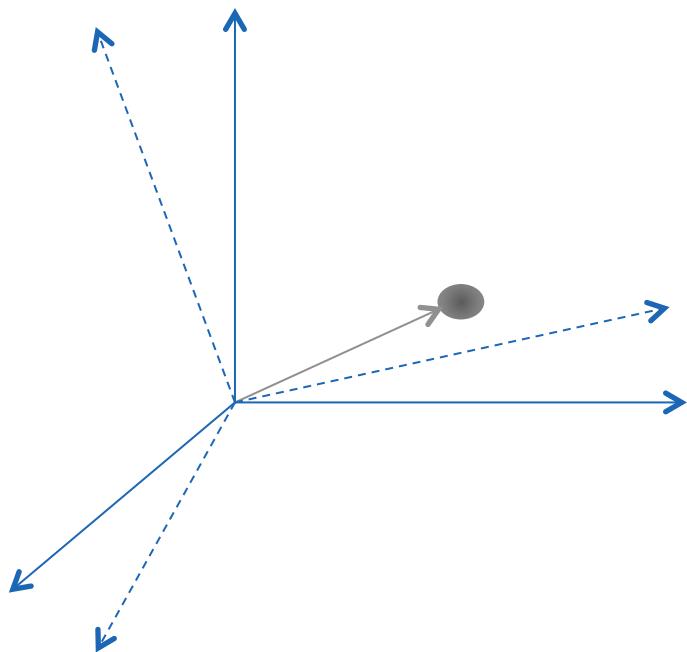
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Basis functions

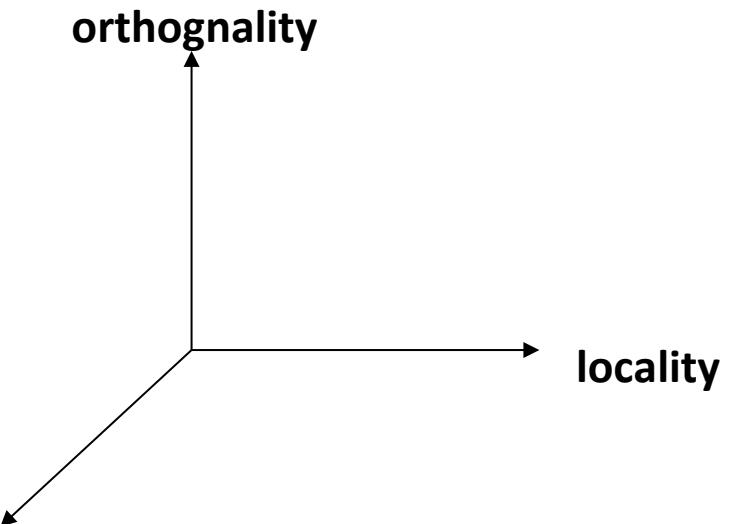


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Attributes of basis functions



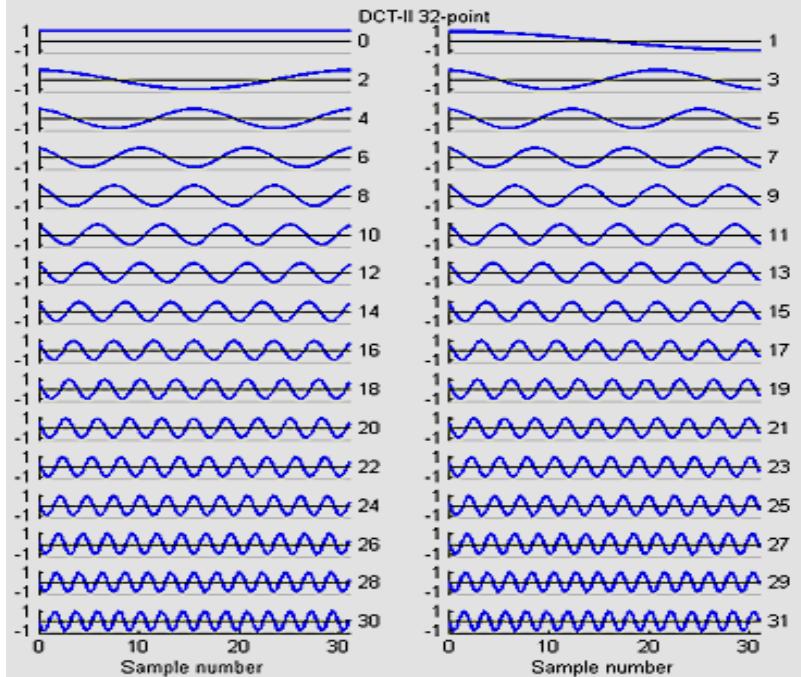
signal dependence





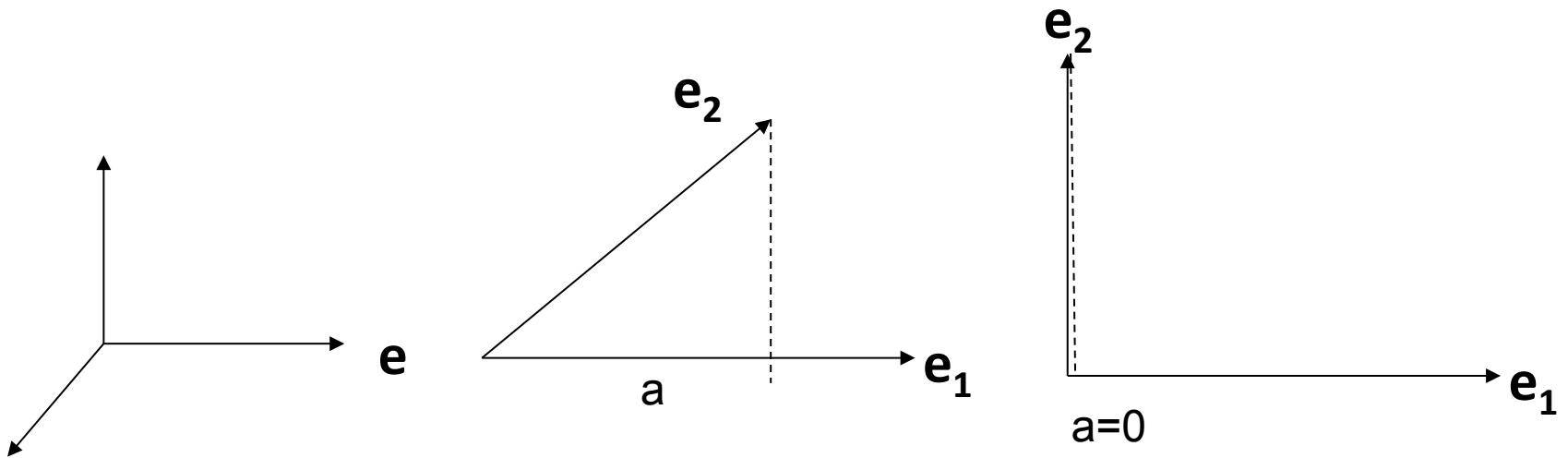
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Fourier Transform



Orthogonality

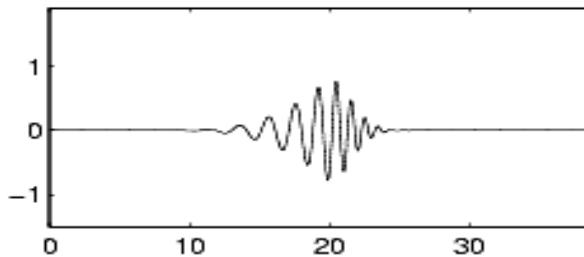
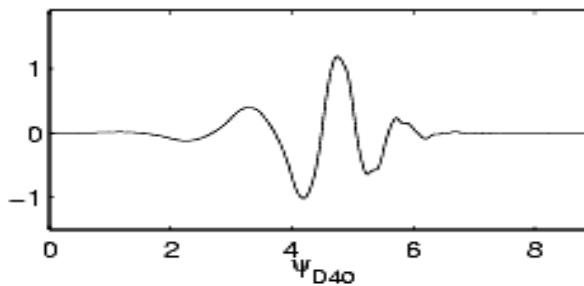
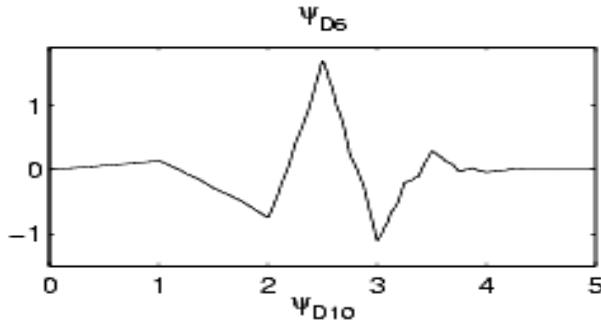
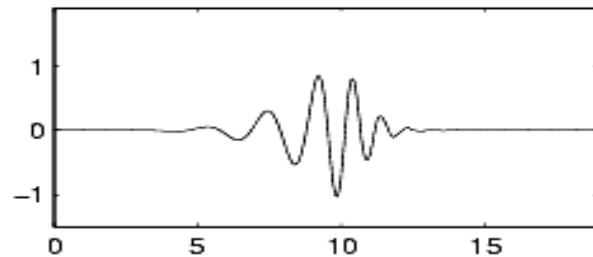
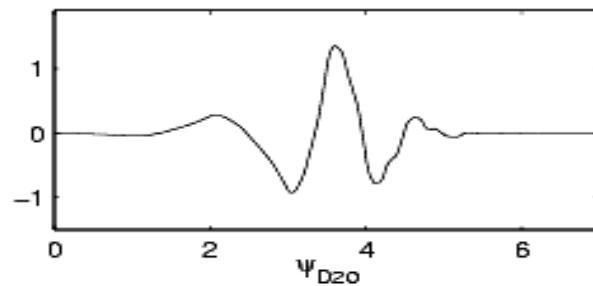
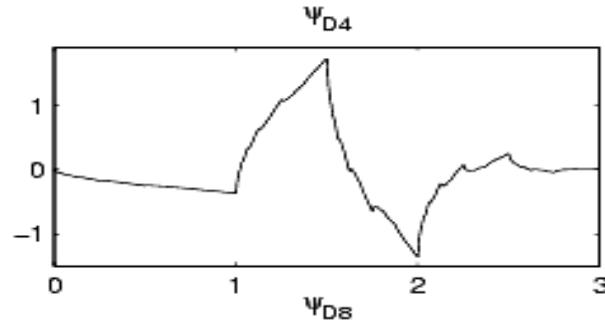
$$x = a_0 \mathbf{e}_0 + a_1 \mathbf{e}_1 + a_2 \mathbf{e}_2 + \dots$$





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Wavelet

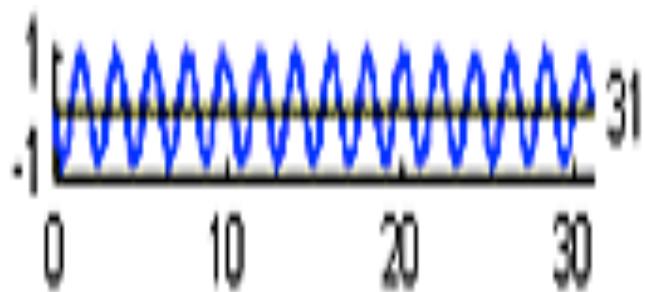
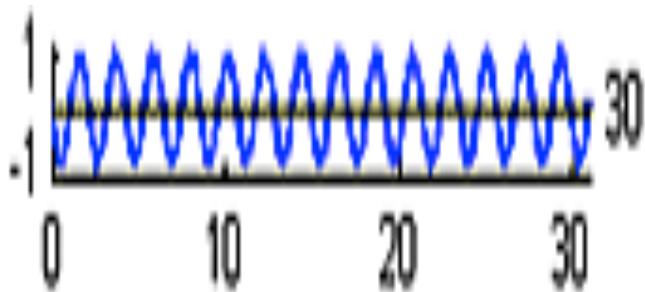




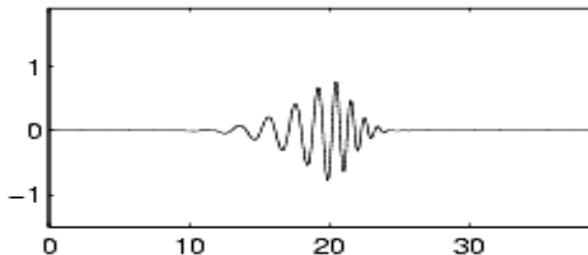
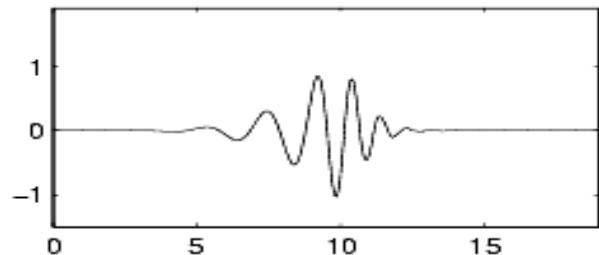
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Locality

Global support



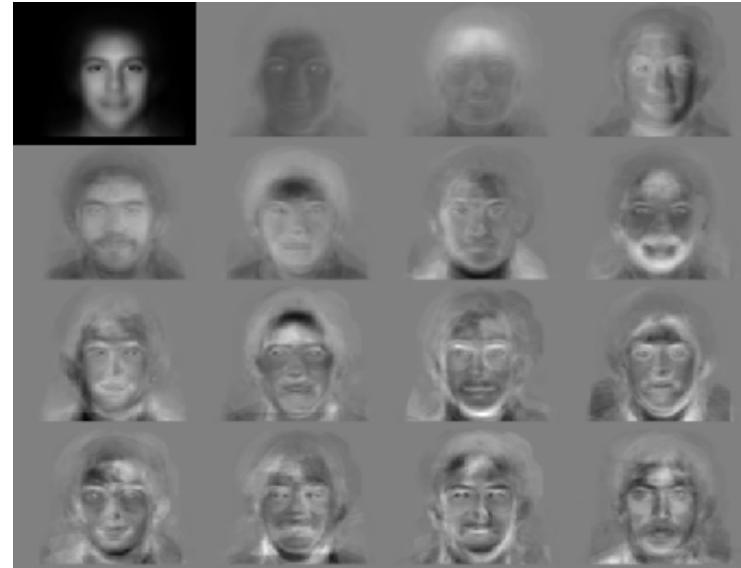
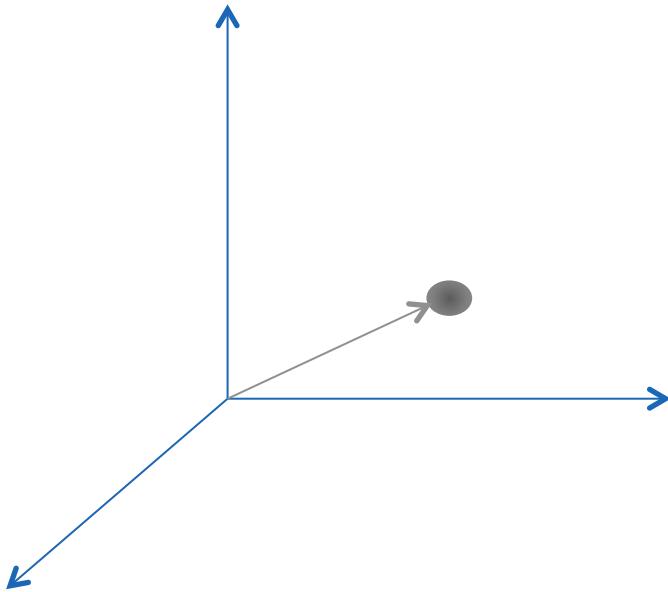
Local support





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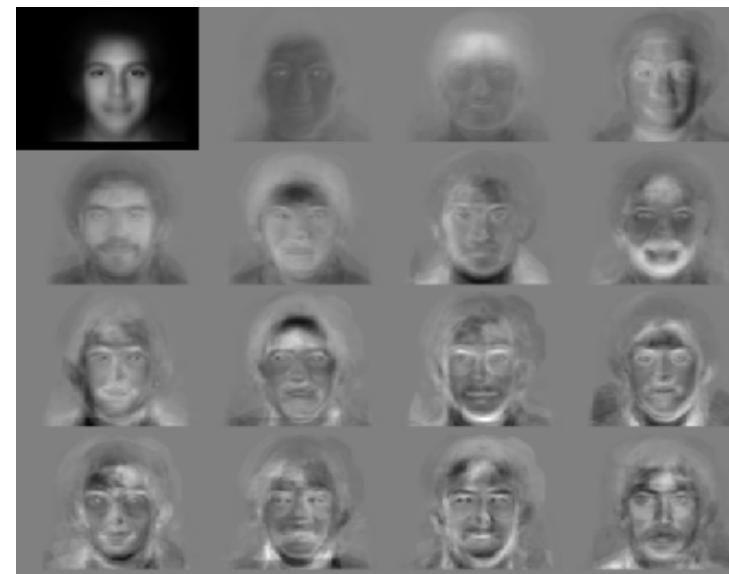
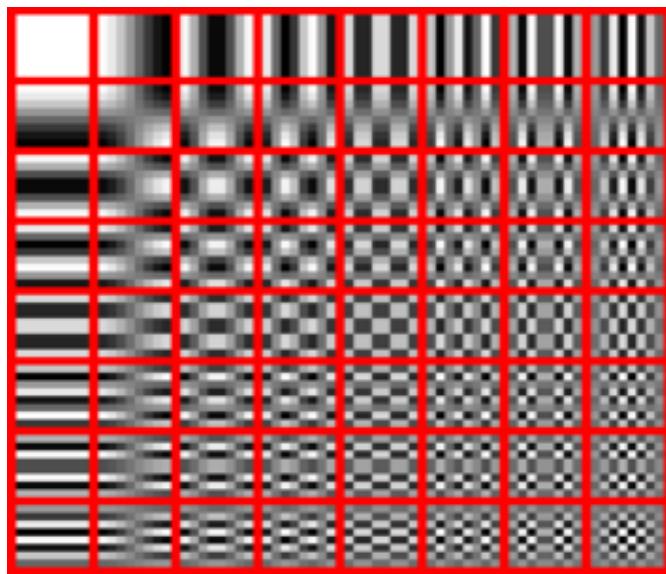
Eigenfaces





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Signal-dependency

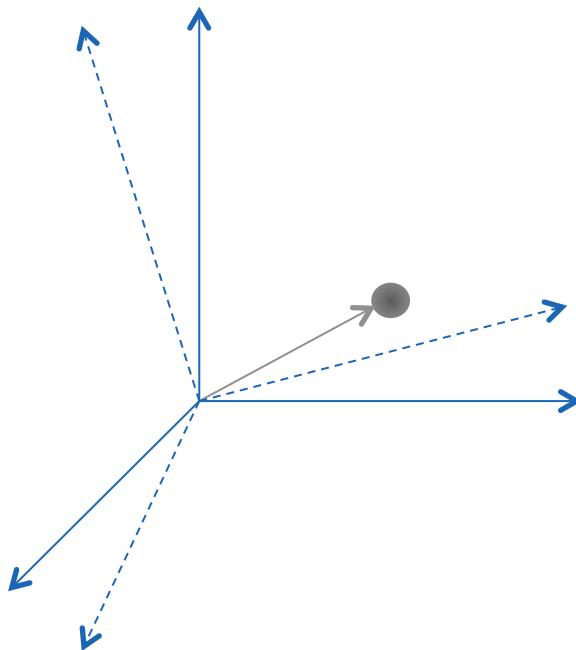




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Computing weights

Signal Representation



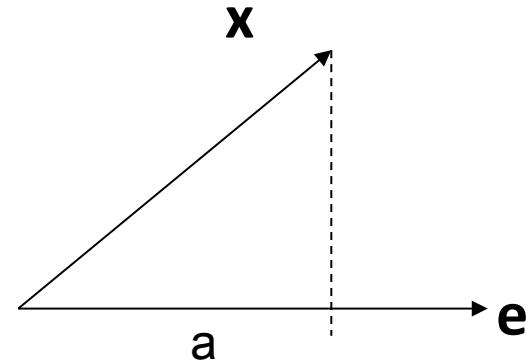
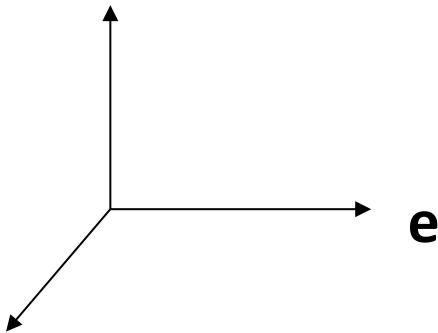
$$x = a_0 \mathbf{e}_0 + a_1 \mathbf{e}_1 + a_2 \mathbf{e}_2 + \dots$$

Two questions

- how to select basis \mathbf{e}_i
- how to calculate a_i

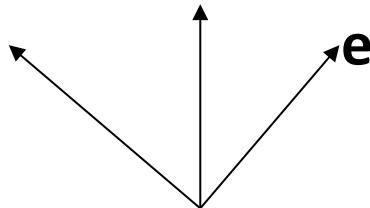
Orthogonal case

$$x = a_0 \mathbf{e}_0 + a_1 \mathbf{e}_1 + a_2 \mathbf{e}_2 + \dots$$



Nonorthogonal case

$$x = a_0 \mathbf{e}_0 + a_1 \mathbf{e}_1 + a_2 \mathbf{e}_2 + \dots$$



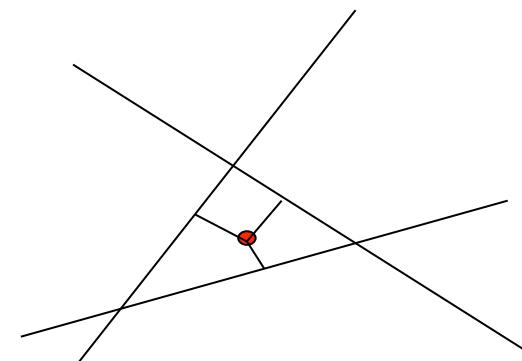
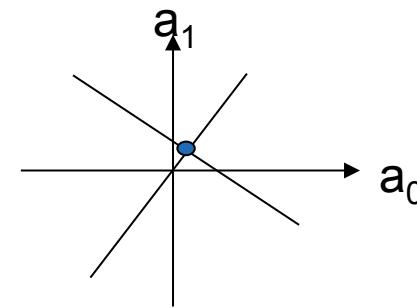
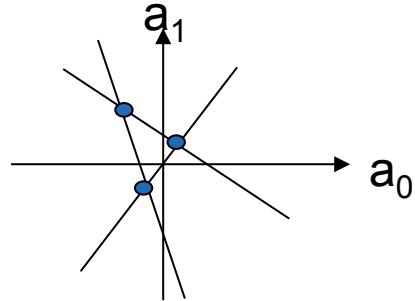
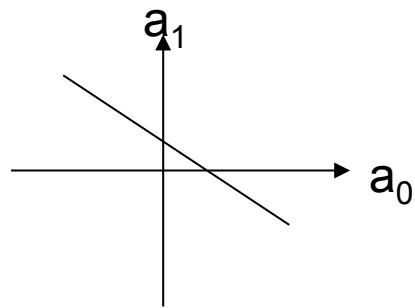
Least-squared (LS) solution!



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Least-Squared Solution

$$x = a_0 \mathbf{e}_0 + a_1 \mathbf{e}_1$$



Least-squared (LS) solution!

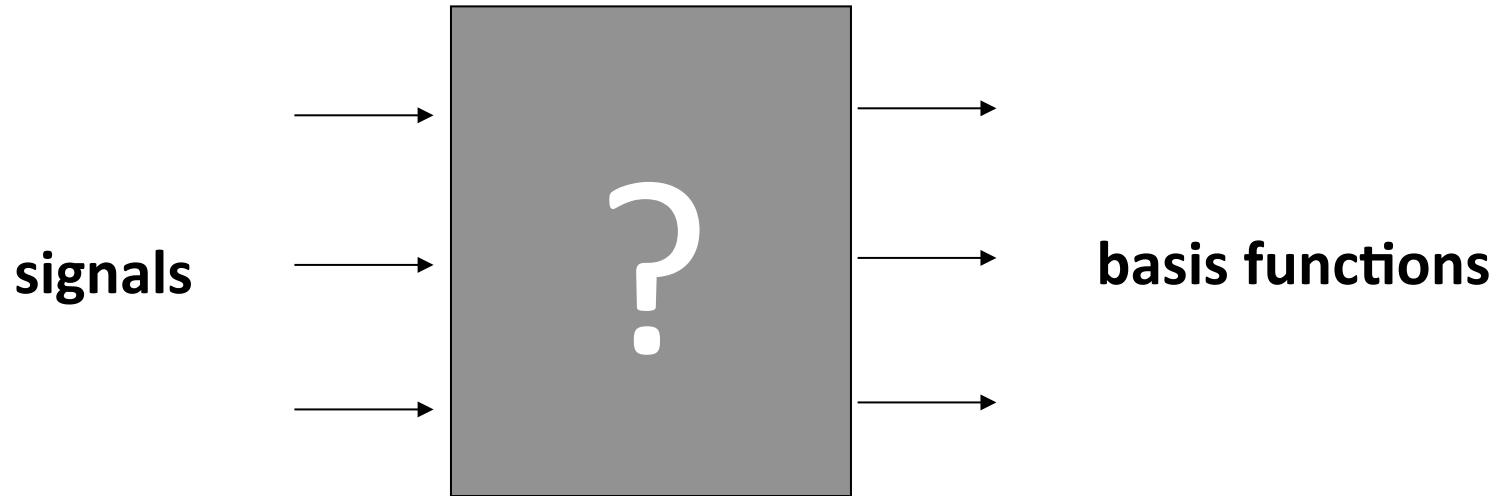


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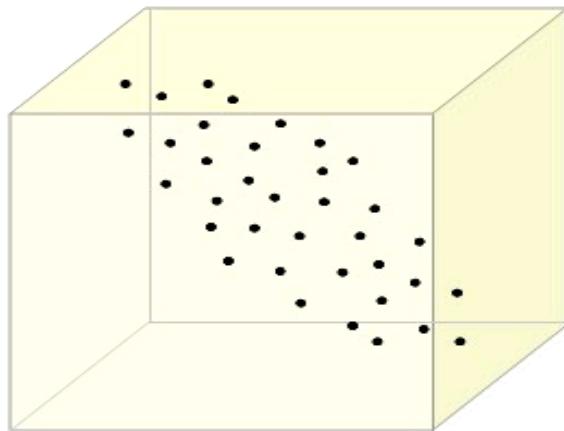
Computing basis functions



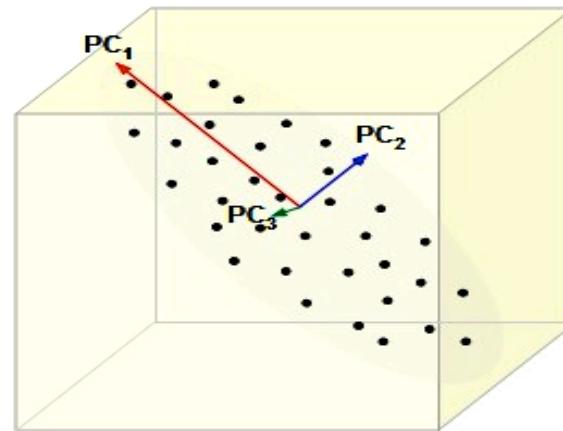
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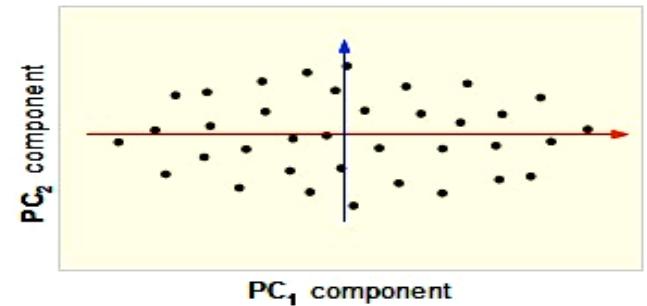
Basis functions from variances



a



b

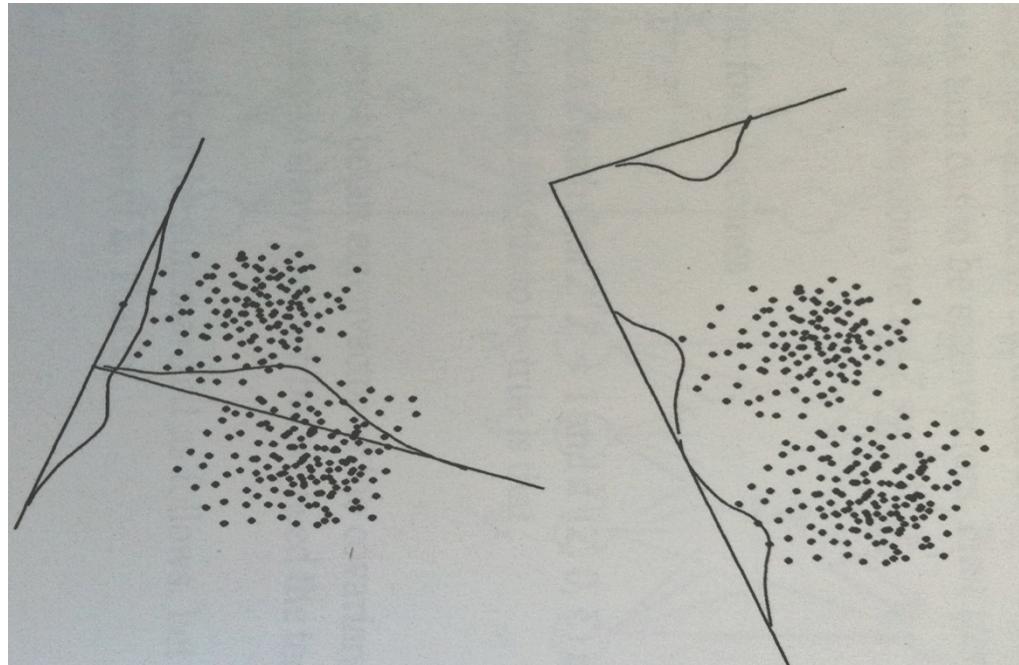


c

Principal Component Analysis (PCA)



Coordinate System





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Basic faces



$$= 0.5 \begin{array}{|c|}\hline \text{Basic Face 1} \\ \hline \end{array} + 0.3 \begin{array}{|c|}\hline \text{Basic Face 2} \\ \hline \end{array} + 0.9 \begin{array}{|c|}\hline \text{Basic Face 3} \\ \hline \end{array} + \dots + 0.1 \begin{array}{|c|}\hline \text{Basic Face 6} \\ \hline \end{array}$$

Surprisingly, the number of basic faces is only around 100!

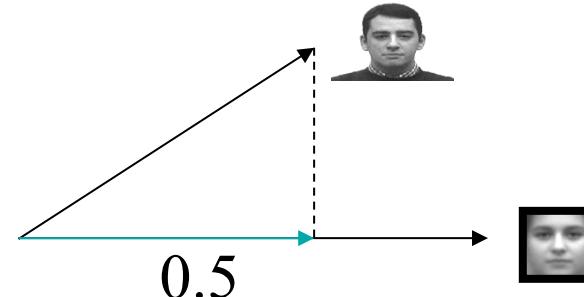


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Face Representation



$$= 0.5 \begin{matrix} \text{[Facial component 1]} \\ \text{[Facial component 1]} \end{matrix} + 0.3 \begin{matrix} \text{[Facial component 2]} \\ \text{[Facial component 2]} \end{matrix} + 0.9 \begin{matrix} \text{[Facial component 3]} \\ \text{[Facial component 3]} \end{matrix} + \dots + 0.1 \begin{matrix} \text{[Facial component 6]} \\ \text{[Facial component 6]} \end{matrix}$$





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Face codes



$$= 0.5 \begin{matrix} \text{[small face image]} \end{matrix} + 0.3 \begin{matrix} \text{[small face image]} \end{matrix} + 0.9 \begin{matrix} \text{[small face image]} \end{matrix} + 0.1 \begin{matrix} \text{[small face image]} \end{matrix}$$



$$= 0.8 \begin{matrix} \text{[small face image]} \end{matrix} + 0.7 \begin{matrix} \text{[small face image]} \end{matrix} + 0.1 \begin{matrix} \text{[small face image]} \end{matrix} + 0.4 \begin{matrix} \text{[small face image]} \end{matrix}$$

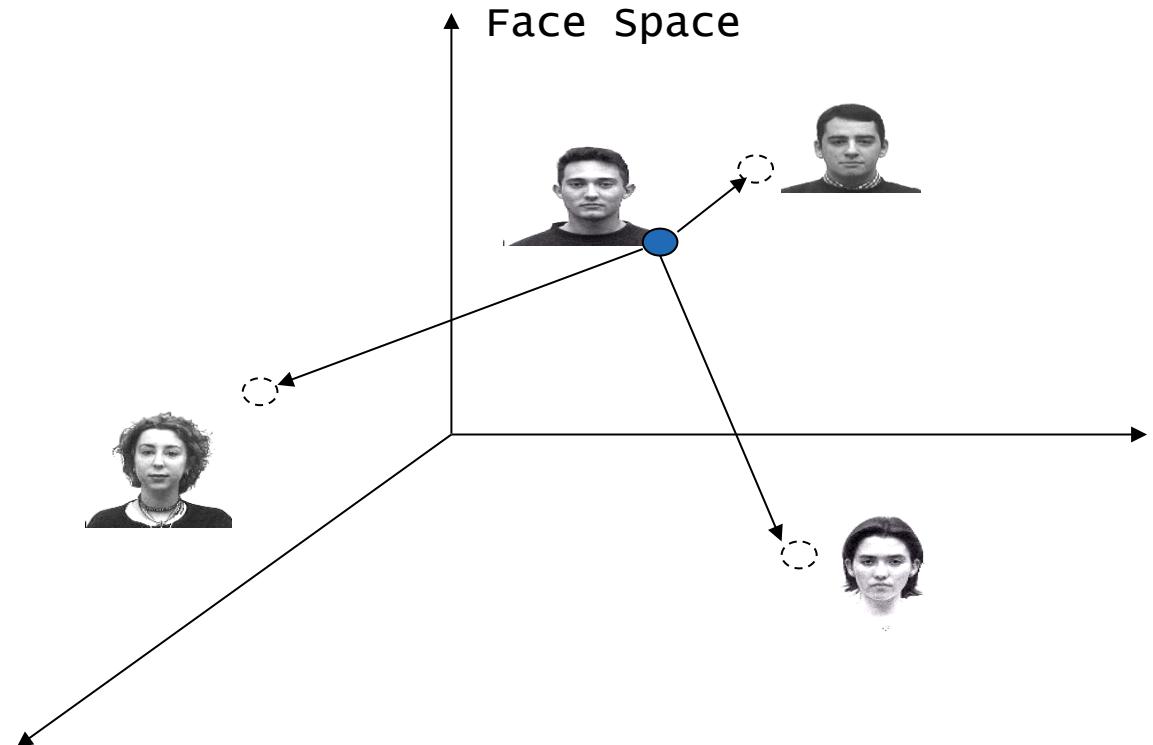


$$= 0.2 \begin{matrix} \text{[small face image]} \end{matrix} + 0.4 \begin{matrix} \text{[small face image]} \end{matrix} + 0.5 \begin{matrix} \text{[small face image]} \end{matrix} + 0.8 \begin{matrix} \text{[small face image]} \end{matrix}$$



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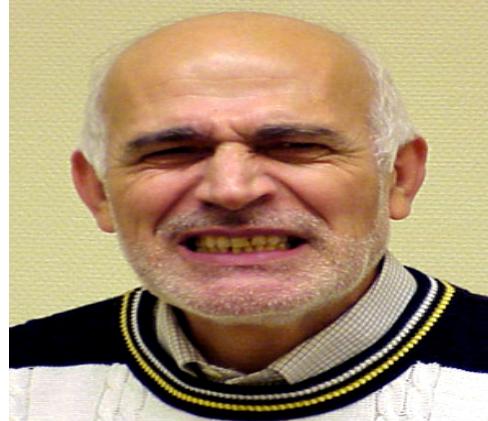
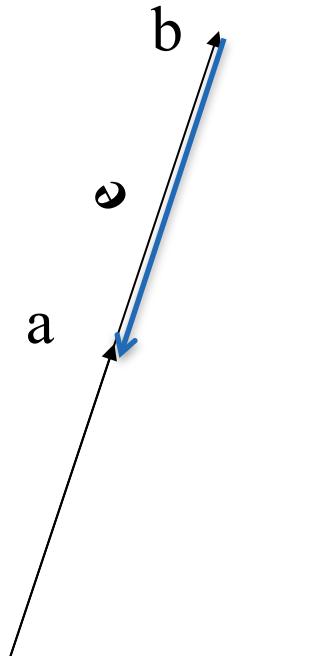
Face recognition = Classification



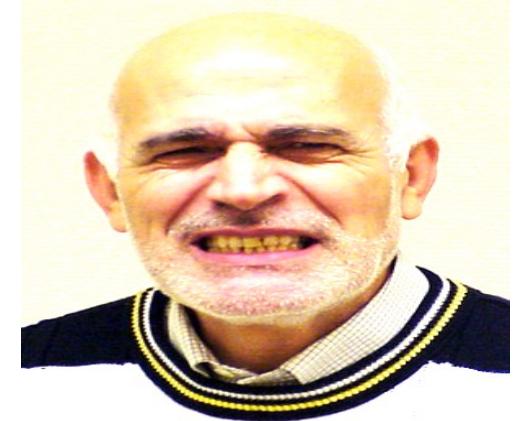


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Vector Distance



a



b

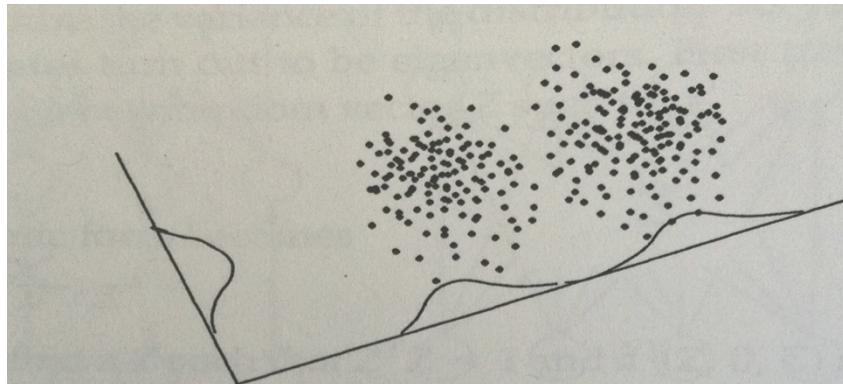
Normalization:

$$\mathbf{a} = x\mathbf{b} + \mathbf{y}$$

Least squared solution

Random Vectors

Consider vectors are drawn from some random distribution that captures the natural variations in the world. A random vector X is specified by a probability density function $p(X)$





- Lab 1

Group 7, Group 9

- Lab2

Group 1, Group 3, Group 9, Group 10, Group 17