## COCEITEC

Central European Institute of Technology
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## Image analysis IV

C9940 3-Dimensional Transmission Electron Microscopy S1007 Doing structural biology with the electron microscope

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## Outline

## Image analysis III

- More on FFTs
- Classification
- Review of multivariate data analysis
- Classification in 2D
- Classification in 3D
- Resolution estimation
- Fourier Shell Correlation
- Expectation value of noise
- "Gold standard" resolution


## Some simple 2D Fourier transforms: a row of points



## Some simple 2D Fourier transforms: a series of lines

## Some simple 2D Fourier transforms:

 a 2D lattice

## Single point



If the point was infinitely sharp, the FFT would be flat.

## Some simple 1D transforms: a sharp point (Dirac delta function)



http://en.labs.wikimedia.org/wiki/Basic_Physics_of_Nuclear_Medicine/Fourier_Methods

## Single point



If the point was infinitely sharp, the FFT would be flat.

## Two points



## Three points



## Five points


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One row


## Two rows



Three rows


Five rows


Full lattice


## Animation




What if?


## Convolution: a review

Adapted from David DeRosier


Set a molecule down at every lattice point.


Cross-correlation: $F^{*}(X) G(X)$


What if?



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$$

What if?


$f(x)$

$g(x)$

$F(X) G(X)$

## Classification

## Reiteration of the problem

## 8 classes of faces, $64 \times 64$ pixels



With noise added

Average:


Before we can average the data, we first should find homogeneous subsets.

Multivariate data analysis (MDA)


# Multivariate data analysis (MDA), or Multivariate statistical analysis (MSA) 

| 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 | 16 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |

Our 16-pixel image can be reorganized into a 16-coordinate vector.

## MDA: Reconstituted images

## Linear combinations of these images will give us approximations of the images that make up the data.



Average Eigenimage \#1 Eigenimage \#2 Eigenimage \#3

Display Select class 1 start key: 1

| * | 3 | * | * | * | \% | * | \% | * | \$ | * | 8 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| \% | * | \% | * | * | * | * | * | * | * | \% | - |
| * | - | 8 | * | * | * | * | - | \% | * | * | $\bigcirc$ |
| * | * | * | - | * | - | * | \% | * | \% | \% | \% |
| \% | \% | \% | $\%$ | \% | * | * | 8 | * | \% | \% | - |
| \% | * | \% | \% | * | 3 | * | * | * | * | * | 9 |
| ) | * | \% | 8 |  |  |  |  |  |  |  |  |

MDA of worm hemoglobin
Average:


## Classification

| 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 | 14 | 15 | 16 |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |



How do we categorize/classify the images?

## K-means classification

A number $K$ of images are chosen as seeds.


BAD: Some clusters may be overrepresented/underrepresented.

## Diday's method of moving centers

Factor 1 vs 2


싱․

Diday's method of moving centers


Diday's method of moving centers


## Diday's method of moving centers



We will note the images that always "travel" together, and will call them a class.

## Dendrogram

CLA/dendrogram.ps


## Dendrogram



## Hierarchical ascendant classification



## Hierarchical Ascendant Classification



All images are represented.
The dendrogram will be too heavily branched to interpret without truncation.

## Binary-tree viewer



BAD: Information about the height of the branch is lost.

## Classification in 3D

# Classification: <br> Reference-based classification vs. Maximum likelihood (ML3D) 

## Reference-based classification: ML3D

- Possible conformations must be - Possible conformations are known.
- The combination of parameters (shift, rotation, class) is chosen from the highest correlation value.
- Possible reference bias not known.

The probability of the occurrence of the parameters (shift, rotation, class) is maximized.

- Random, data-dependent

RELION is a variation of maximum likelihood.

## Seeding ML3D classification

We split the data set into $K$ classes at random.


There will be slight differences in the reconstructions.
We will iteratively maximize the likelihood of a particle belonging to a particular class.

## How good is our reconstruction?

## How do we evaluate the quality of a reconstruction?

We split the data set into halves and compare them.


## Fourier Shell Correlation (FSC)




Properties:

- Fourier terms have amplitude + phase.
- Correlation values range from -1 to +1 .
- Noise should give an average of 0 .
- The comparison is done as a function of spatial frequency (or "resolution")


## Fourier Shell Correlation curve



## FSC curve with expectation value of noise



## Why does $\sigma$ vary with spatial frequency?




Random walks:
Why signal-to-noise improves with $\sqrt{ } N$

## The "Drunkard's walk"



Let's conduct an experiment.

## The "Drunkard's walk"



We're going to assume that each step is random and independent of previous steps.

## The "Drunkard's walk"


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## The teetotaler's walk

|  |  |  |  |  |  |  |  |  |
| :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- | :--- |
| -4 | -3 | -2 | -1 | 0 | 1 | 2 | 3 | 4 |

$t=2$
$t=3$
$t=4$

## Expectation value



The expected distance that "noise" travels increases with $\sqrt{ } \mathrm{N}$. However, it is not as fast as the distance that "signal" travels.
Thus, as we collect more data, the SNR increase by $N / \sqrt{N}=\sqrt{ } N$

## Random walks: more information



## Expectation values

and how they related to resolution criteria

## With small N , behavior is more unpredictable



One resolution criterion was to compare the FSC to, say, $3^{*} \sigma$. BUT:
The $\sigma$ value describes the behavior of unaligned noise.

## Review: model bias



The model bias can yields false correlations in real space is equivalent to false correlations in Fourier space.

## Refinement: classical and "gold standard"



## Different resolution criteria



## Thank you for your attention

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