



FFTs in Graphics and Vision

Conclusion

Announcements

Assignment 3 due May 9th.





Key Ideas

Functions Are Vectors

We can think of functions as (complex) vectors in a high-dimensional space.

Representations

Translations and rotations of functions can be thought of as linear, norm-preserving, transformations.

Irreducible Representations

To understand how translations and rotations act on functions, we strive to decompose the space of functions into the smallest sub-spaces in which these transformations are contained.



Essential Facts

Schur's Lemma (Corollary)

If the group is commutative, the irreducible representations are all one-dimensional.

Homogenous Polynomials

Homogenous polynomials of fixed degree are sub-representations.

Self-Adjoint Operators

If a linear operator is self-adjoint (symmetric) there is an orthogonal basis of eigenvectors.

Commuting Self-Adjoint Operators

The spaces of eigenvectors of the operator with the same eigenvalue are sub-representation.

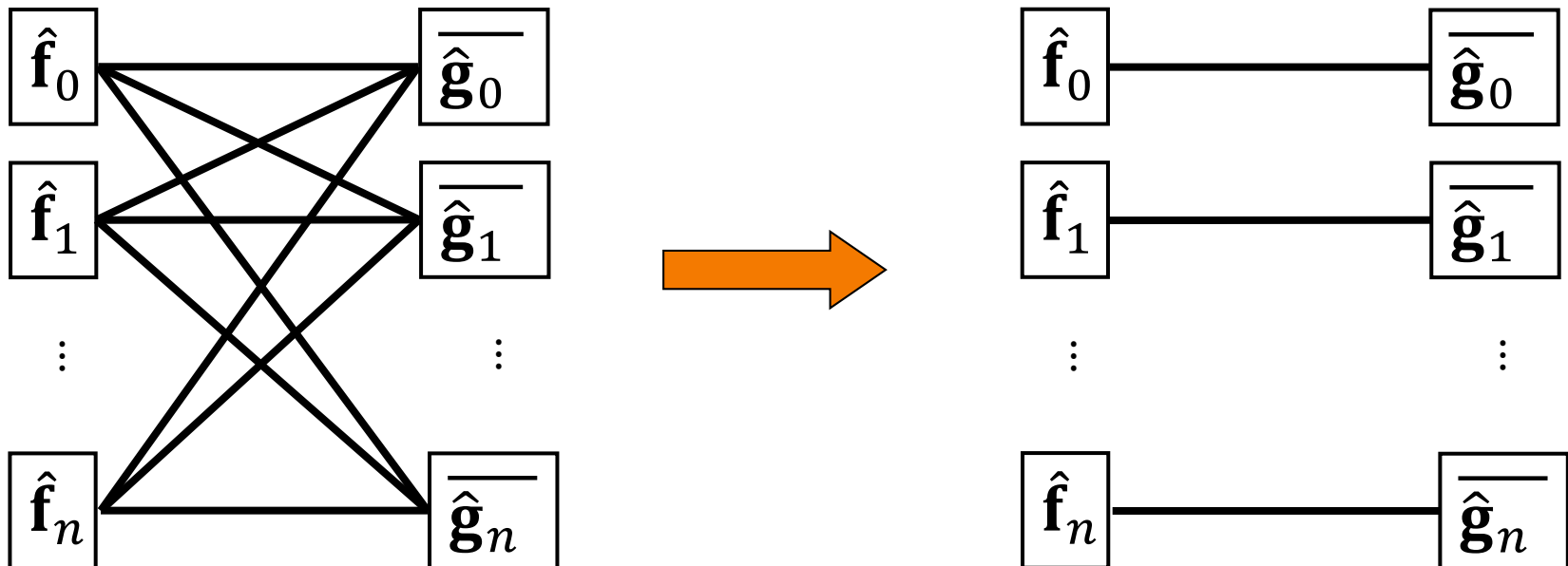


Implications / Applications

Fast Correlation:

For correlation, we only need to perform cross-multiplication of coefficients in the same frequency:

$$D_{f,g}(\alpha) = \sum_k \hat{\mathbf{f}}_k \cdot \overline{\hat{\mathbf{g}}_k} \cdot D_k(\alpha)$$





Implications / Applications

Fast Correlation:

For correlation, we only need to perform cross-multiplication of coefficients in the same frequency.

This reduces the implementation of correlation to:

1. Computing a forward frequency transform
2. Doing the intra-frequency cross multiplication
3. Computing an inverse frequency transform



Implications / Applications

Fast Correlation:

For correlation, we only need to perform cross-multiplication of coefficients in the same frequency.

For Translational Correlation:

- | | |
|-------------------------------|---------------|
| 1. Computing the FFT: | $O(N \log N)$ |
| 2. Doing the multiplication: | $O(N)$ |
| 3. Computing the inverse FFT: | $O(N \log N)$ |

Brute force would have been $O(N^2)$.

Optimal would be $O(N)$.



Implications / Applications

Fast Correlation:

For correlation, we only need to perform cross-multiplication of coefficients in the same frequency.

For Rotational Correlation:

- | | |
|-------------------------------|-------------------|
| 1. Computing the SHT: | $O(N^2 \log^2 N)$ |
| 2. Doing the multiplication: | $O(N^3)$ |
| 3. Computing the inverse WDT: | $O(N^3 \log^2 N)$ |

Brute force would have been $O(N^5)$.

Optimal would have been $O(N^3)$.



Implications / Applications

Fast Correlation (Applications):

For image processing, we have seen applications of correlation in:

- Filtering



Edge Detection

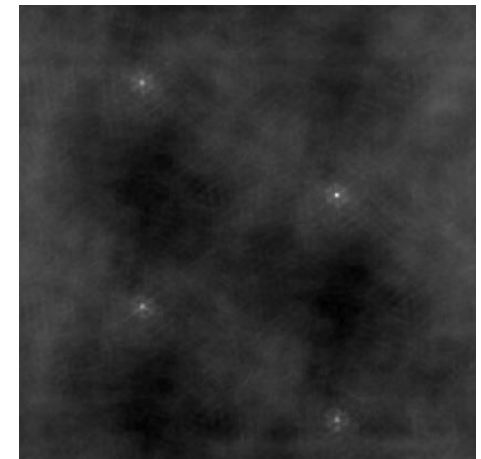
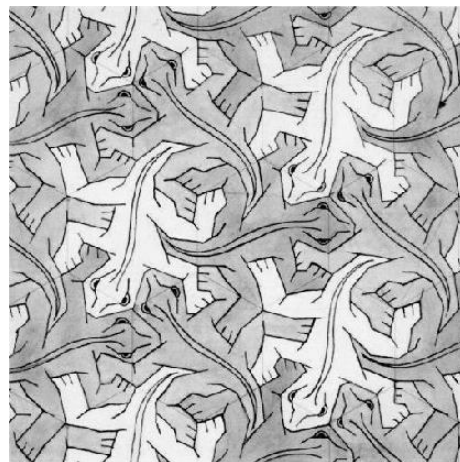


Implications / Applications

Fast Correlation (Applications):

For image processing, we have seen applications of correlation in:

- Filtering
- Pattern Recognition



Template Matching

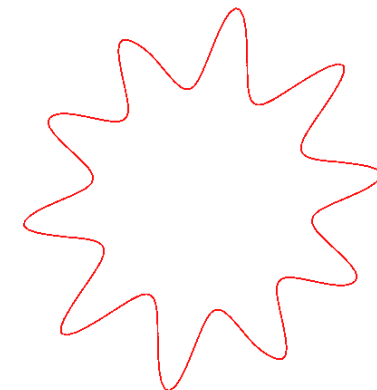
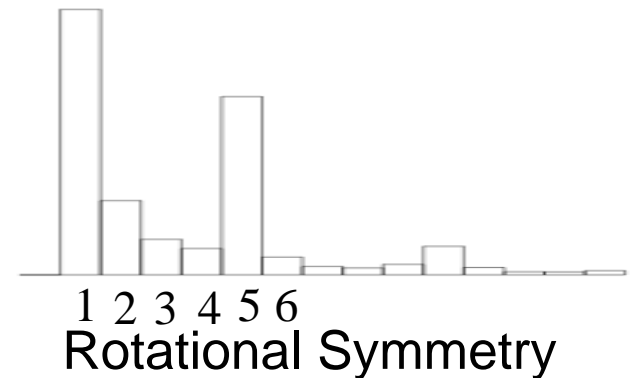


Implications / Applications

Fast Correlation (Applications):

For image processing, we have seen applications of correlation in:

- Filtering
- Pattern Recognition
- Symmetry Detection



Reflective Symmetry



Implications / Applications

Fast Correlation (Applications):

We have even seen some unexpected applications resulting from the fact that the Laplacian is a symmetric operator commuting with translation in:

- Solving PDEs



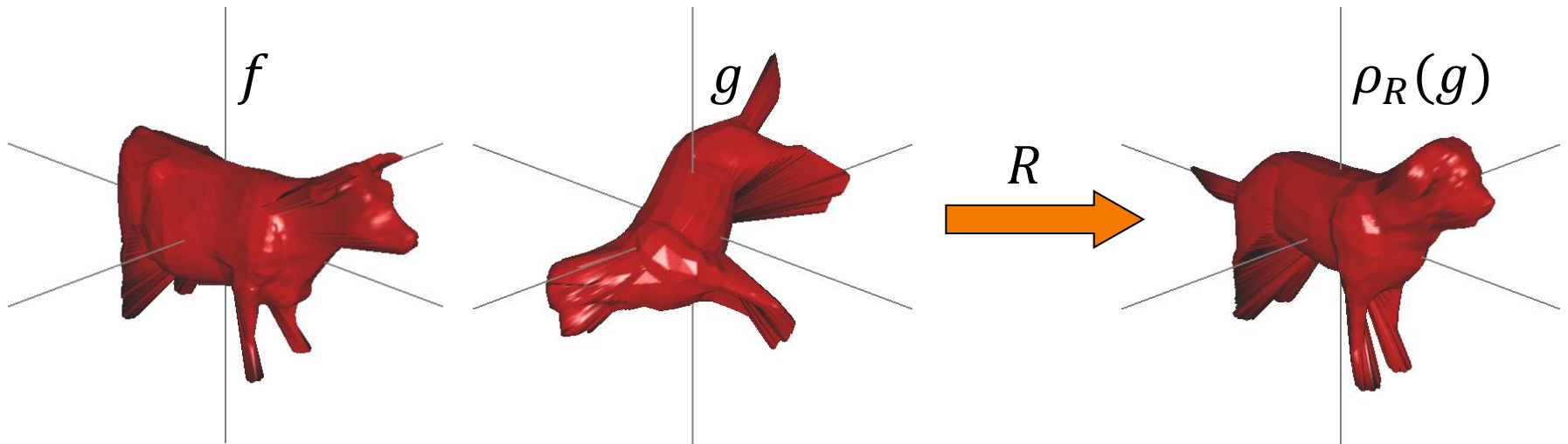


Implications / Applications

Fast Correlation (Applications):

For the group of rotations in 3D, we have seen applications in:

- Shape Alignment



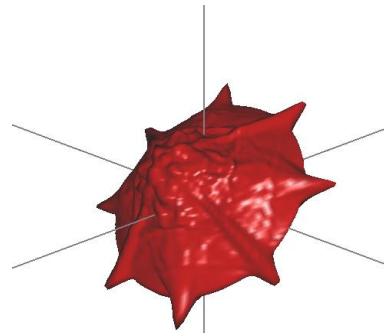
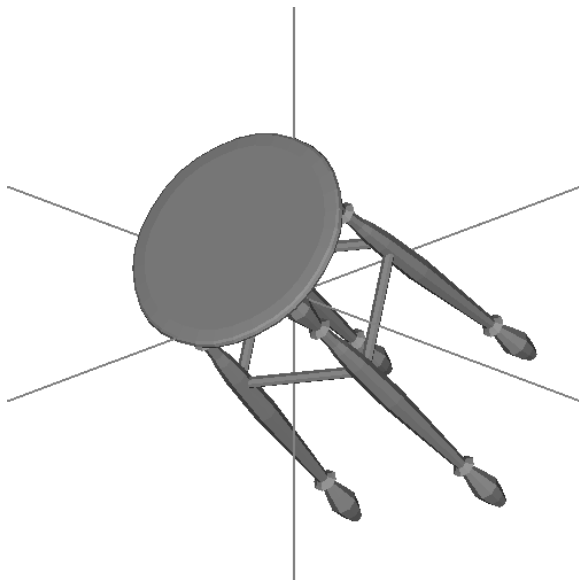


Implications / Applications

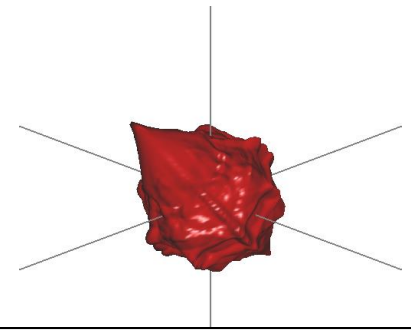
Fast Correlation (Applications):

For the group of rotations in 3D, we have seen applications in:

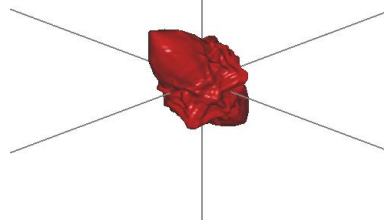
- Shape Alignment
- Symmetry Detection



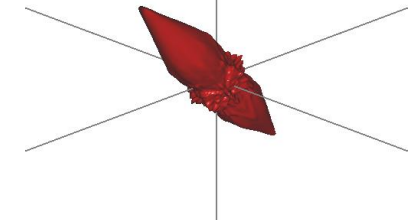
Reflective Symmetries



2-Fold
Rotational Symmetries



3-Fold
Rotational Symmetries



4-Fold
Rotational Symmetries



Thank You!

