

# FFTs in Graphics and Vision

Conclusion

### **Announcements**



Assignment 3 due May 6<sup>th</sup>.

## **Key Ideas**



#### **Functions Are Vectors**

We can think of functions as (complex) vectors in a highdimensional 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 subrepresentations.

### **Self-Adjoint Operators**

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

### Commuting Self-Adjoint Operators

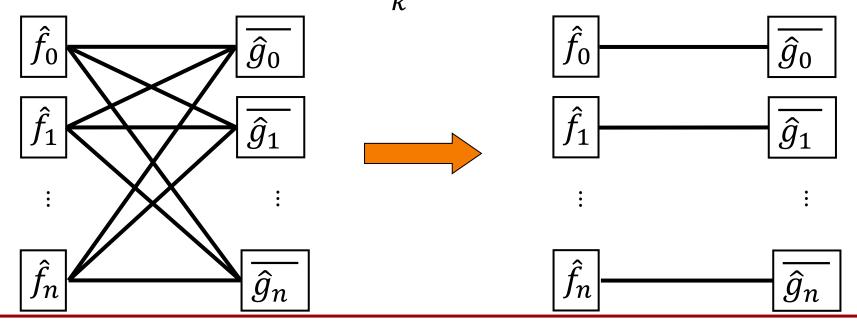
The spaces of eigenvectors of the operator with the same eigenvalue are sub-representation.



### **Fast Correlation:**

For correlation, we only need to perform crossmultiplication of coefficients in the same frequency:

$$D_{f,g}(\alpha) = \sum_{k} \hat{f}_k \cdot \overline{\hat{g}_k} \cdot \mathbf{D}_k(\alpha)$$





#### **Fast Correlation:**

For correlation, we only need to perform crossmultiplication of coefficients in the same frequency:

$$D_{f,g}(\theta,\phi,\psi) = \sum_{l} \sum_{m=-l} \hat{f}_{lm} \cdot \overline{\hat{g}_{lm}} \cdot \mathbf{D}_{l}^{mm'}(\theta,\phi,\psi)$$



#### **Fast Correlation:**

For correlation, we only need to perform crossmultiplication 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



#### **Fast Correlation:**

For correlation, we only need to perform crossmultiplication 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).



#### **Fast Correlation:**

For correlation, we only need to perform crossmultiplication 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  $O(N^5)$ .

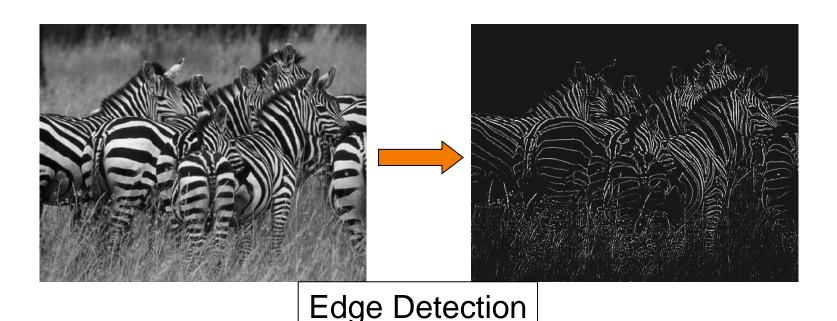
Optimal would have been  $O(N^3)$ .



### Fast Correlation (Applications):

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

Filtering



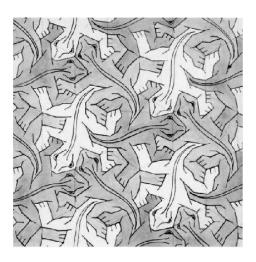


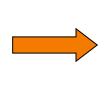
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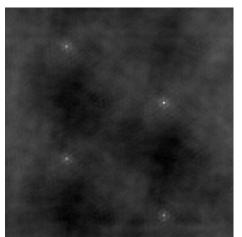
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- Filtering
- Pattern Recognition









**Template Matching** 

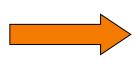


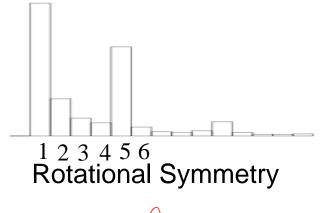
### Fast Correlation (Applications):

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

- Filtering
- Pattern Recognition
- Symmetry Detection







Reflective Symmetry



### 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

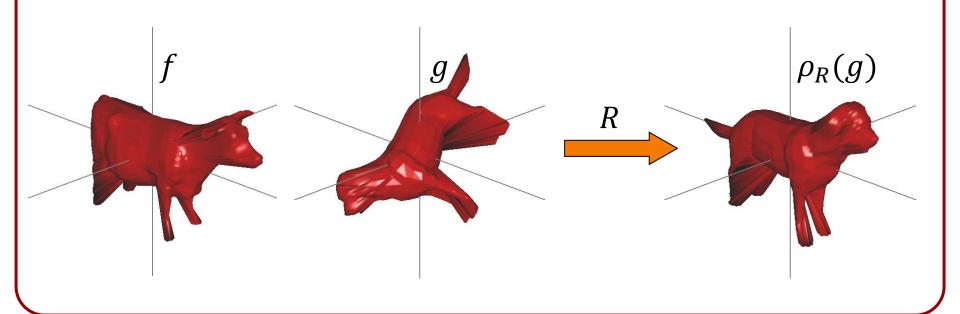




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For the group of rotations in 3D, we have seen applications in:

Shape Alignment





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

- Shape Alignment
- Symmetry Detection

