



# **FFTs in Graphics and Vision**

Spherical Convolution  
and  
Axial Symmetry Detection

# Announcements

Assignment 3 has been posted!





# Outline

- Math Review
  - Symmetry
  - General Convolution
- Spherical Convolution
- Axial Symmetry Detection



# Math Review

## Symmetry:

Given a unitary representation of a group  $G$  on a vector space  $V$ , we say that a vector  $v \in V$  is invariant under the action of  $G$  if for all  $g \in G$ :

$$\rho_g(v) = v$$

The set of  $G$ -invariant vector  $V_G$  is a vector space.



# Math Review

## Symmetry:

The linear map  $\pi_G$  is a projection onto  $V_G$ , if:

- $\pi_G(v) \in V_G$  for all  $v \in V$
- $\pi_G(v) = v$  for all  $v \in V_G$
- $\langle v, w - \pi_G(w) \rangle = 0$  for all  $v \in V_G, w \in V$ .

The map  $\pi_G$  is the map sending a vector  $v$  to the closest  $G$ -invariant vector.



# Math Review

## Symmetry:

The measure of symmetry of a vector  $v$  with respect to the group  $G$  is the size of its projection onto the space of  $G$ -invariant vectors:

$$\text{Sym}^2(v, G) = \|\pi_G(v)\|^2$$



# Math Review

## Convolution:

Given functions  $f(p)$  and  $g(p)$ , the convolution of the two functions is defined as:

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$



# Math Review

## Convolution:

If we hold the function  $f$  fixed we get a map from the space of functions back into itself:

$$C_f(g) = f * g$$

## Claim:

The map  $C_f$  is a linear operator.





# Math Review

## Convolution:

If we hold the function  $f$  fixed we get a map from the space of functions back into itself:

$$C_f(g) = f * g$$

## Claim:

Given functions  $f$  and  $h$  and scalars  $\alpha$  and  $\beta$ :

$$\begin{aligned} C_f(\alpha g + \beta h)(q) &= \int f(q - p) \cdot (\alpha \cdot g(p) + \beta \cdot h(p)) dp \\ &= \alpha \int f(q - p) \cdot g(p) dp + \beta \int f(q - p) \cdot h(p) dp \\ &= \alpha \cdot C_f(g) + \beta \cdot C_f(h) \end{aligned}$$



# Math Review

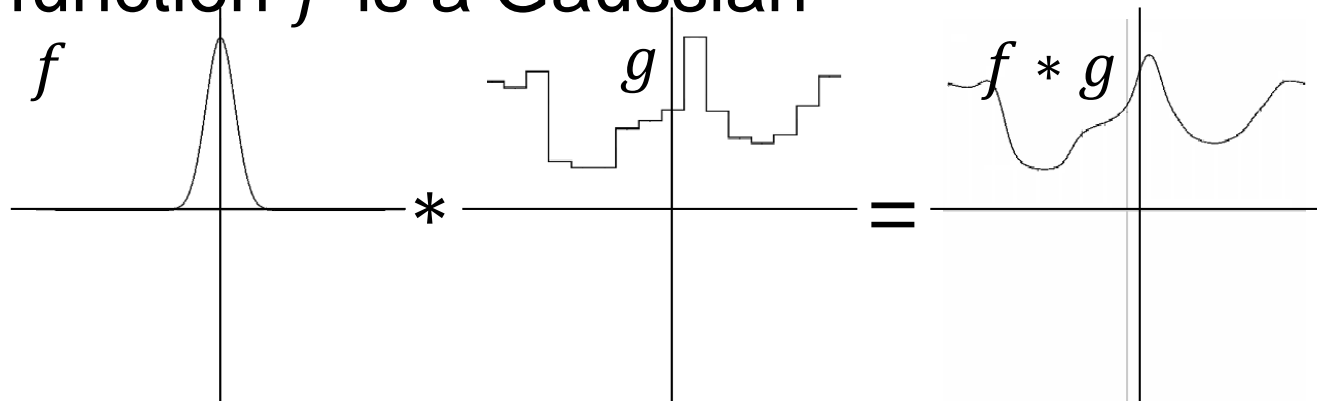
## Convolution:

Assume that the function  $f$  is real-valued and radial, i.e. the value of  $f$  at a point  $p$  is completely determined by the distance of  $p$  from the origin:

$$f(p) = \tilde{f}(|p|)$$

## Example:

The function  $f$  is a Gaussian





# Math Review

## Convolution:

Assume that the function  $f$  is real-valued and radial, i.e. the value of  $f$  at a point  $p$  is completely determined by the distance of  $p$  from the origin:

$$f(p) = \tilde{f}(|p|)$$

## Claim:

In this case,  $C_f$  is self-adjoint (i.e. symmetric).



# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

We need to show that for any functions  $g$  and  $h$ :

$$\langle C_f(g), h \rangle = \langle g, C_f(h) \rangle$$

Expanding the left side, we get:

$$\langle C_f(g), h \rangle = \int (C_f(g))(p) \cdot \overline{h(p)} dp$$



# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

$$\langle C_f(g), h \rangle = \int (C_f(g))(p) \cdot \overline{h(p)} dp$$

Writing out the operator  $C_f$ , we get:

$$\langle C_f(g), h \rangle = \int (f * g)(p) \cdot \overline{h(p)} dp$$



# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

$$\langle C_f(g), h \rangle = \int (f * g)(p) \cdot \overline{h(p)} dp$$

Expressing the convolution as an integral gives:

$$\langle C_f(g), h \rangle = \int \left( \int f(p - q) \cdot g(q) dq \right) \cdot \overline{h(p)} dp$$



# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

$$\langle C_f(g), h \rangle = \int \left( \int f(p - q) \cdot g(q) dq \right) \cdot \overline{h(p)} dp$$

Changing the order of integration, we get:

$$\langle C_f(g), h \rangle = \int \int f(p - q) \cdot g(q) \cdot \overline{h(p)} dp dq$$



# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

$$\langle C_f(g), h \rangle = \int \int f(p - q) \cdot g(q) \cdot \overline{h(p)} dp dq$$

Using the fact that  $f$  is real-valued and radial:

$$\langle C_f(g), h \rangle = \int \int g(q) \cdot \overline{f(q - p) \cdot h(p)} dp dq$$





# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

$$\langle C_f(g), h \rangle = \int \int g(q) \cdot \overline{f(q - p) \cdot h(p)} dp dq$$

Moving the integration inside:

$$\langle C_f(g), h \rangle = \int g(q) \left( \int \overline{f(q - p) \cdot h(p)} dp \right) dq$$



# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

$$\langle C_f(g), h \rangle = \int g(q) \left( \int \overline{f(q - p) \cdot h(p)} dp \right) dq$$

Using the equation for convolution, we get:

$$\langle C_f(g), h \rangle = \int g(q) \cdot \overline{(f * h)(q)} dq$$



# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

$$\langle C_f(g), h \rangle = \int g(q) \cdot \overline{(f * h)(q)} dq$$

Using the equation for  $C_f$ , we get:

$$\langle C_f(g), h \rangle = \int f(q) \cdot \overline{(C_f(h))(q)} dq$$



# Math Review

$$C_f(g) = f * g$$

$$(f * g)(q) = \int f(q - p) \cdot g(p) dp$$

Proof:

$$\langle C_f(g), h \rangle = \int g(q) \cdot \overline{(C_f(h))(q)} dq$$

And finally, using the equation for the dot-product:

$$\langle C_f(g), h \rangle = \langle f, C_f(h) \rangle$$

# Outline



- Math Review
- Spherical Convolution
- Axial Symmetry Detection

# Spherical Convolution/Correlation



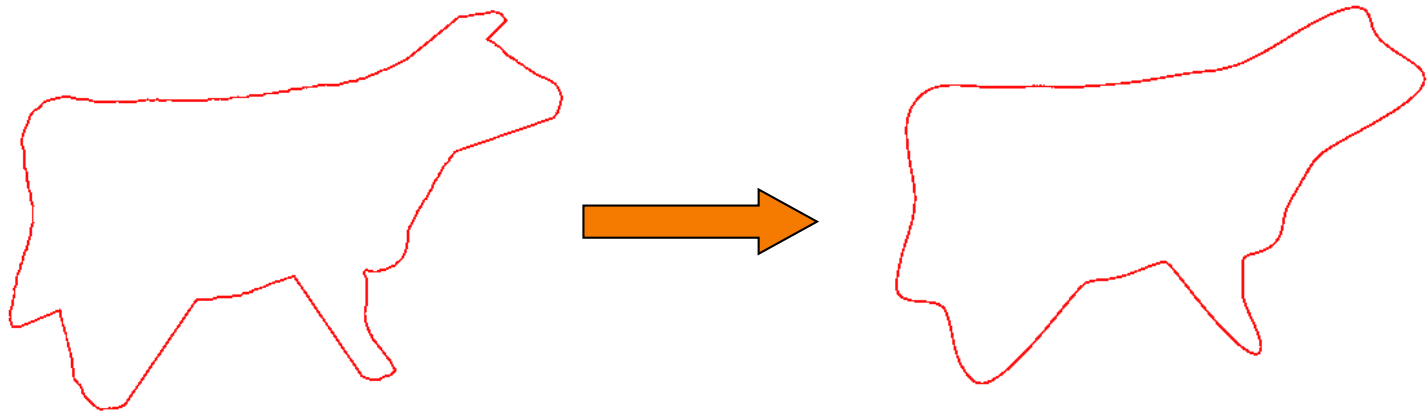
In the case of the circle we used convolution / correlation for two different tasks:

# Spherical Convolution/Correlation



In the case of the circle we used convolution / correlation for two different tasks:

1. We used convolution for operations like smoothing

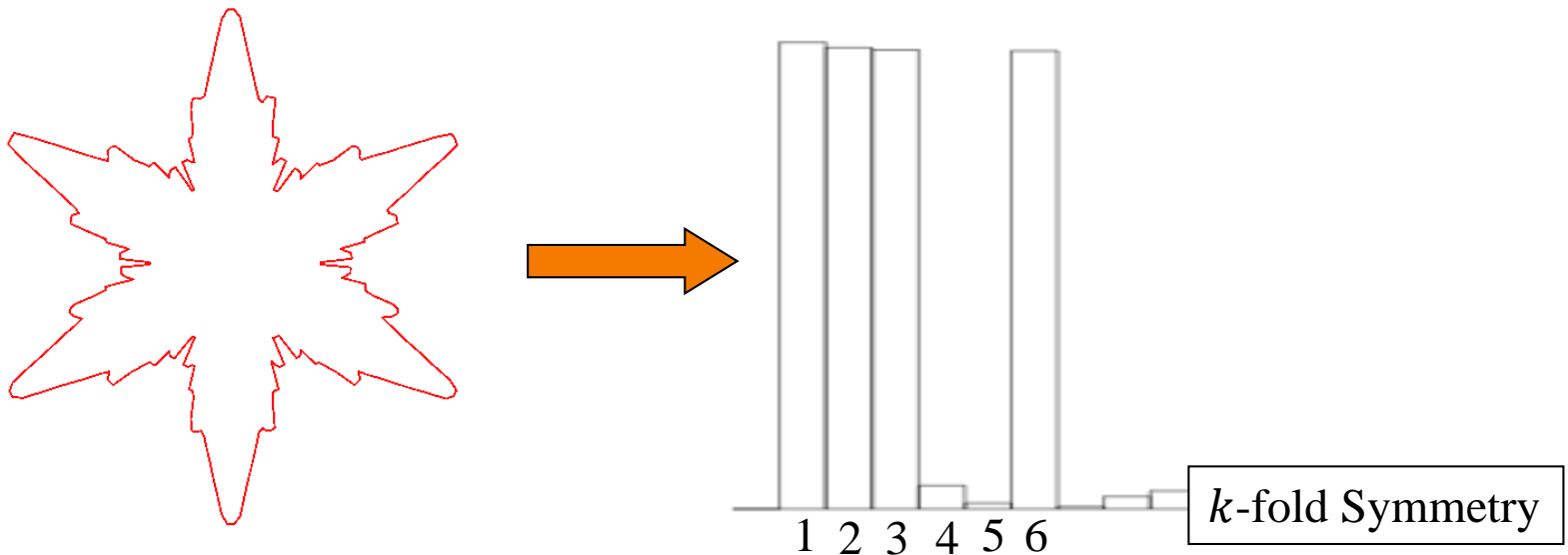


# Spherical Convolution/Correlation



In the case of the circle we used convolution / correlation for two different tasks:

1. We used convolution for operations like smoothing
2. We used correlation for operations like alignment and symmetry detection





# Spherical Convolution/Correlation



Up to now, we thought of these two operations as essentially the same.

The situation changes as we move to functions on a sphere.



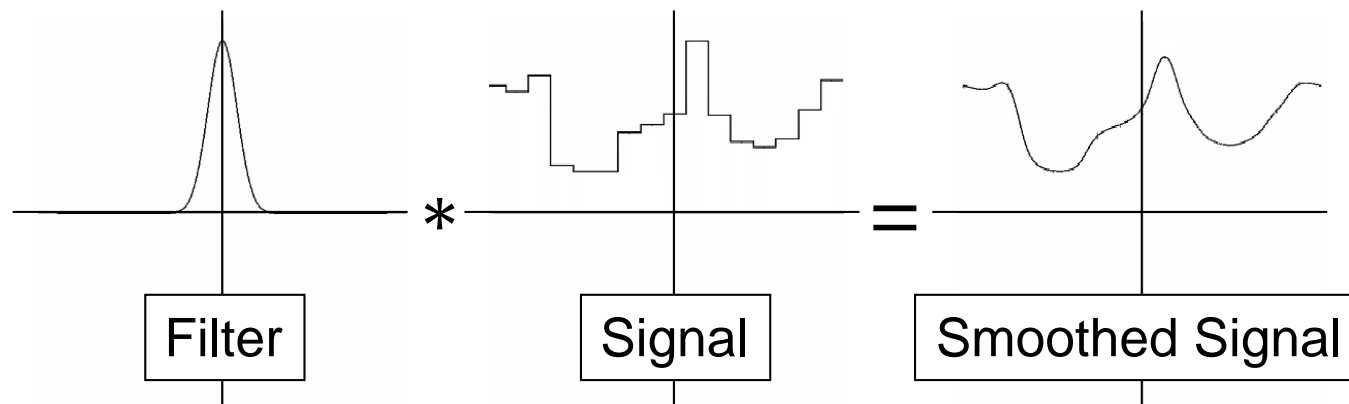
# Spherical Convolution/Correlation

When we perform an operation like smoothing, the input is:

- A function on the circle defining the signal, and
- A function on the circle defining the smoothing filter

The output of the operation is:

- A function on the circle



# Spherical Convolution/Correlation

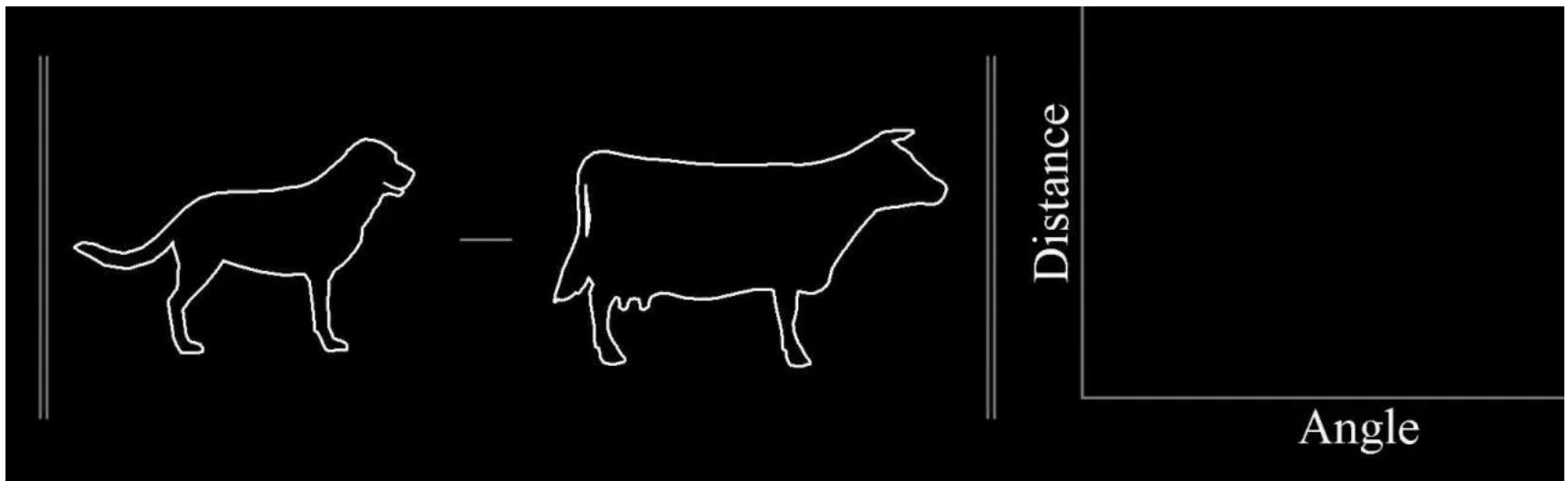


When we perform an operation like alignment, the input is:

- Two functions on a circle

The output is:

- A function on the space of 2D rotations



# Spherical Convolution/Correlation



In the case of a circle, the situation is simpler because the space of rotations is itself a circle:

There is a one-to-one mapping from points on a circle to rotations, with a point on a circle with angle  $\theta$  corresponding to a rotation by an angle of  $\theta$ .

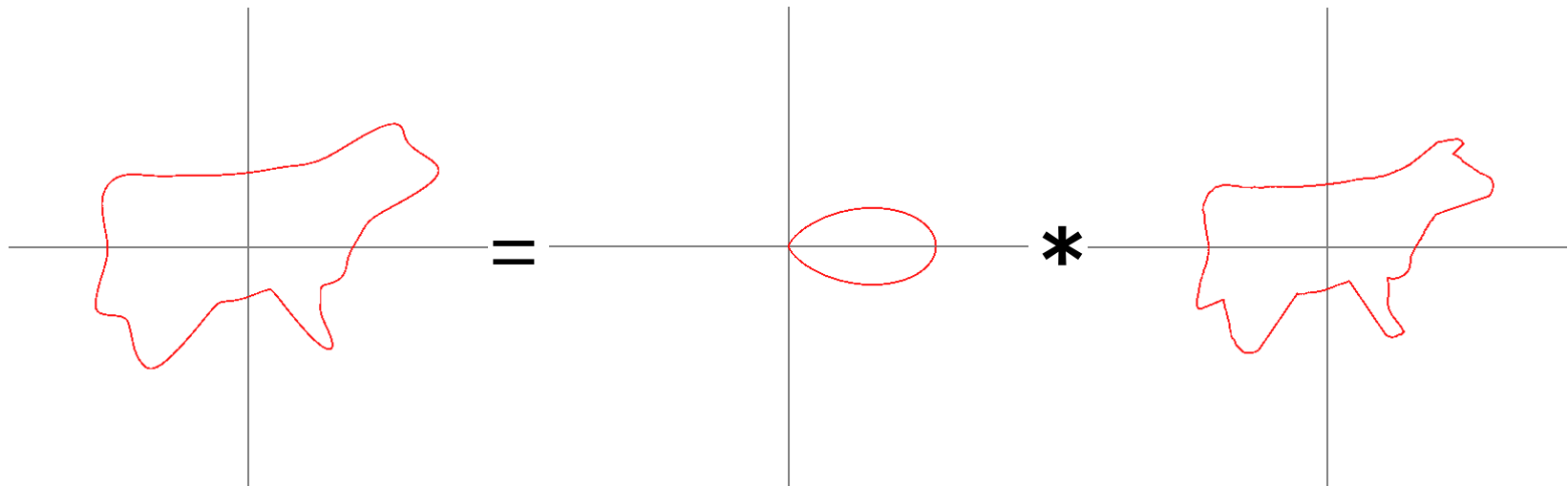
In the case of the sphere, the situation becomes more complicated:

The sphere is a 2D space while the rotations are a 3D space, so there can't be a one-to-one mapping.



# Spherical Convolution

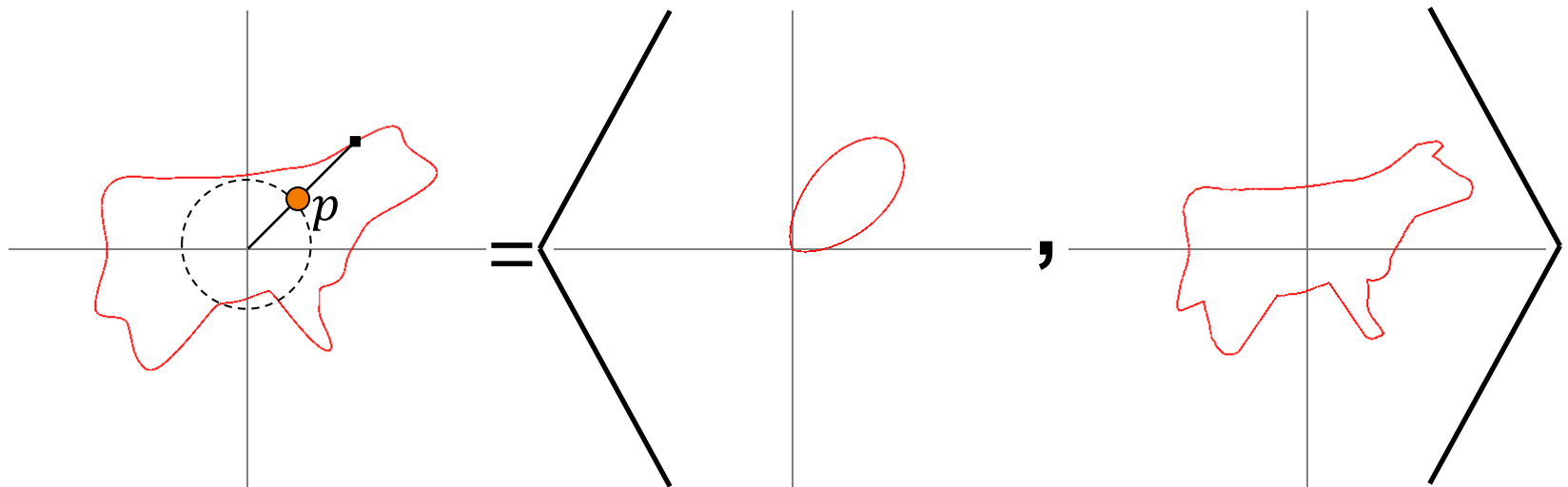
In the case of a circle, we compute the value of the smoothed function at  $p$  by rotating the filter so that  $(1,0)$  maps to  $p$  and then we compute the inner product of the signal with the rotated filter.





# Spherical Convolution

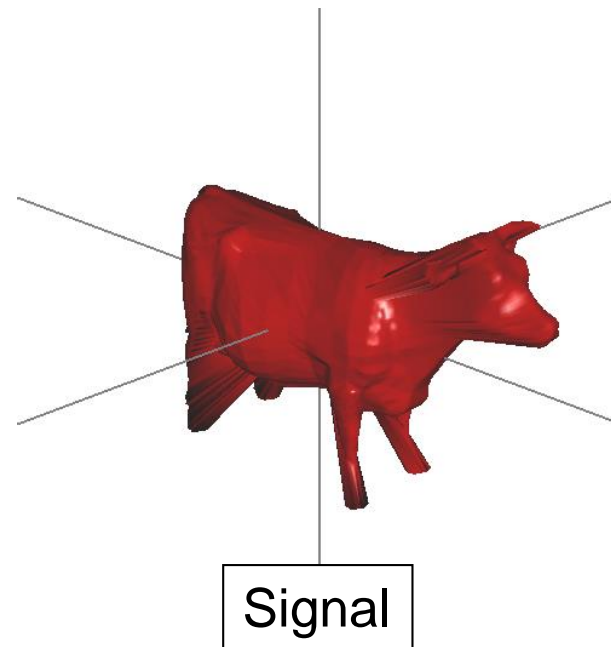
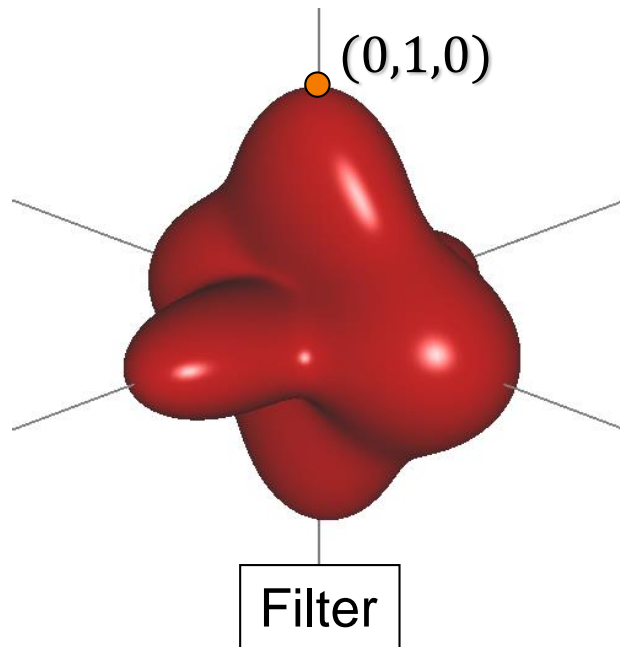
In the case of a circle, we compute the value of the smoothed function at  $p$  by rotating the filter so that  $(1,0)$  maps to  $p$  and then we compute the inner product of the signal with the rotated filter.





# Spherical Convolution

We can try and apply the same type of approach to the case of spherical functions.

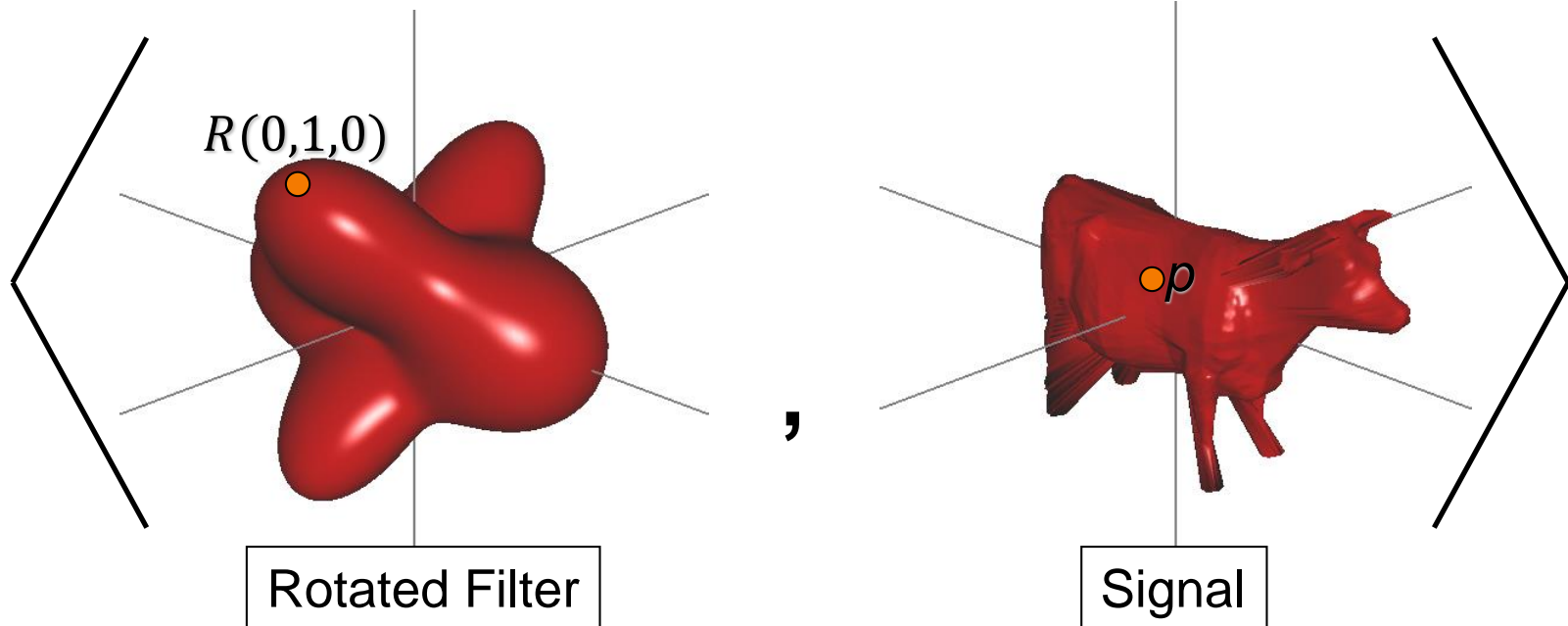




# Spherical Convolution

We would like to define a new function on the sphere whose value at the point  $p$  is obtained by:

Finding a rotation  $R$  that maps the North pole to  $p$  and then compute the inner product of the signal with the rotated filter.

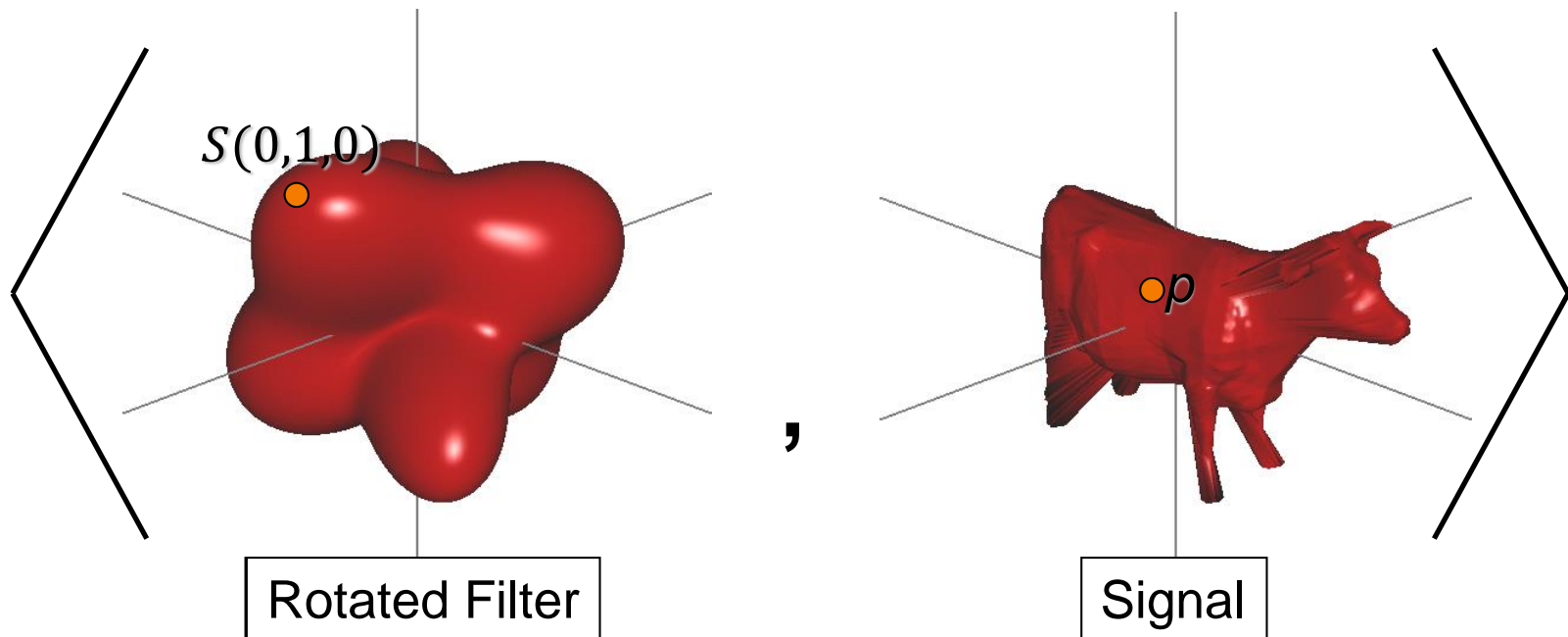






# Spherical Convolution

The problem is that there are many different rotations that send the North pole to the point  $p$ , so this does not lead to a well-defined notion of smoothing.



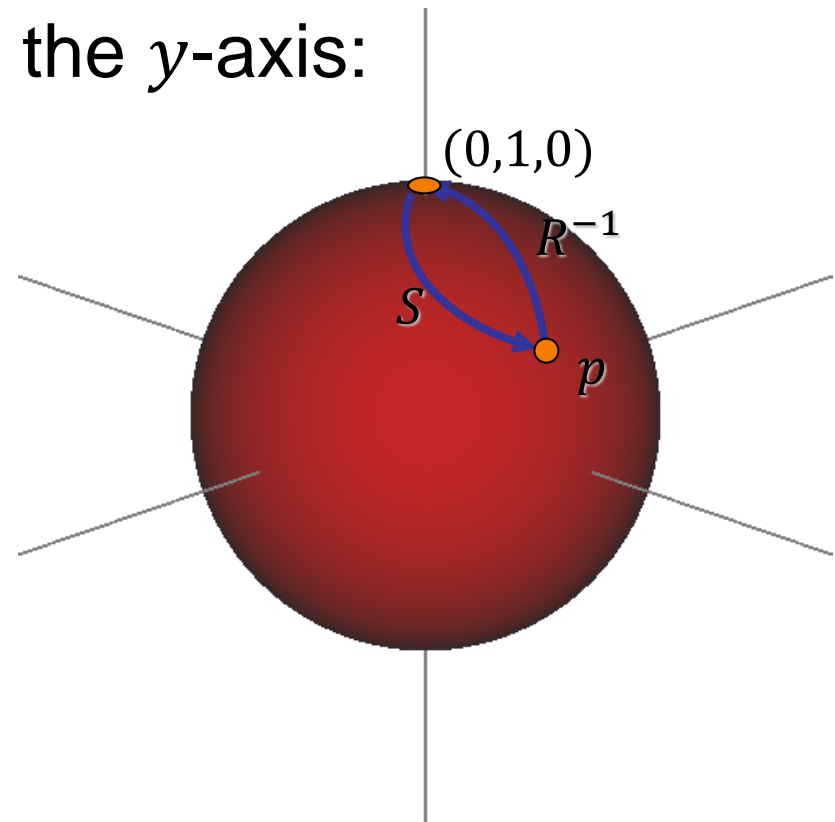


# Spherical Convolution

Recall:

If we have two rotations  $R$  and  $S$  mapping the North pole to the point  $p$ , the rotations must differ by an initial rotation about the  $y$ -axis:

$$S = R \cdot R_y(\psi)$$





# Spherical Convolution

Recall:

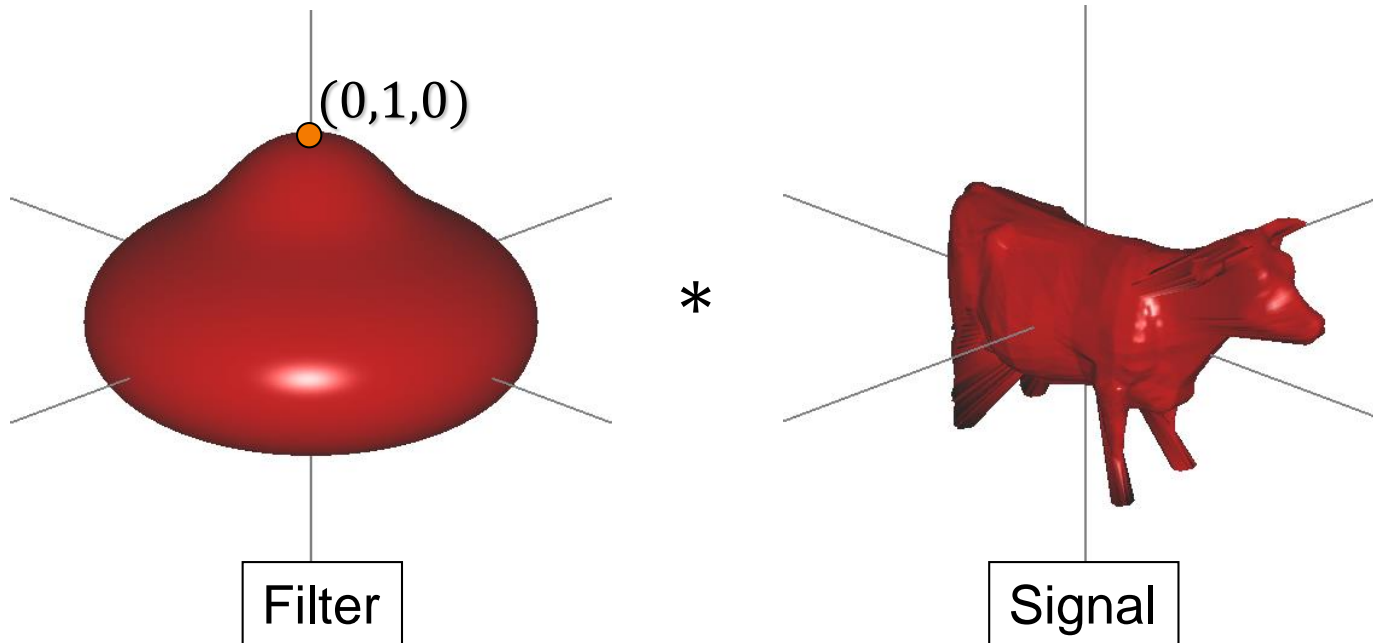
Thus, we can make the notion of smoothing well-defined by ensuring that the initial rotation about the  $y$ -axis does not change the filter.



# Spherical Convolution

Recall:

This means that we can extend the circular notion of smoothing to the sphere if we ensure that the filter is symmetric about the  $y$ -axis:

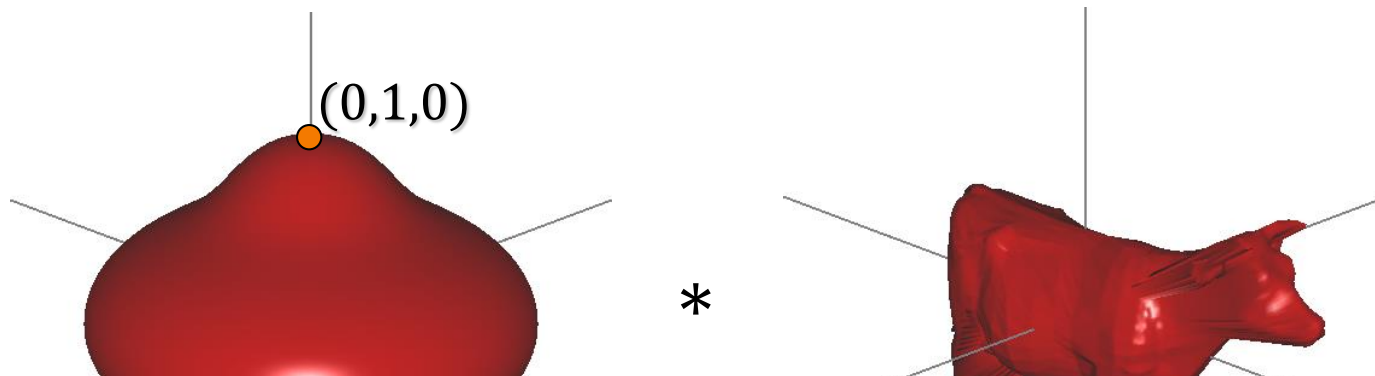




# Spherical Convolution

Recall:

This means that we can extend the circular notion of smoothing to the sphere if we ensure that the filter is symmetric about the  $y$ -axis:



If  $R$  and  $S$  are rotations mapping the North pole to  $p$ , then the rotation of the filter by either  $R$  or  $S$  will give the same spherical function!

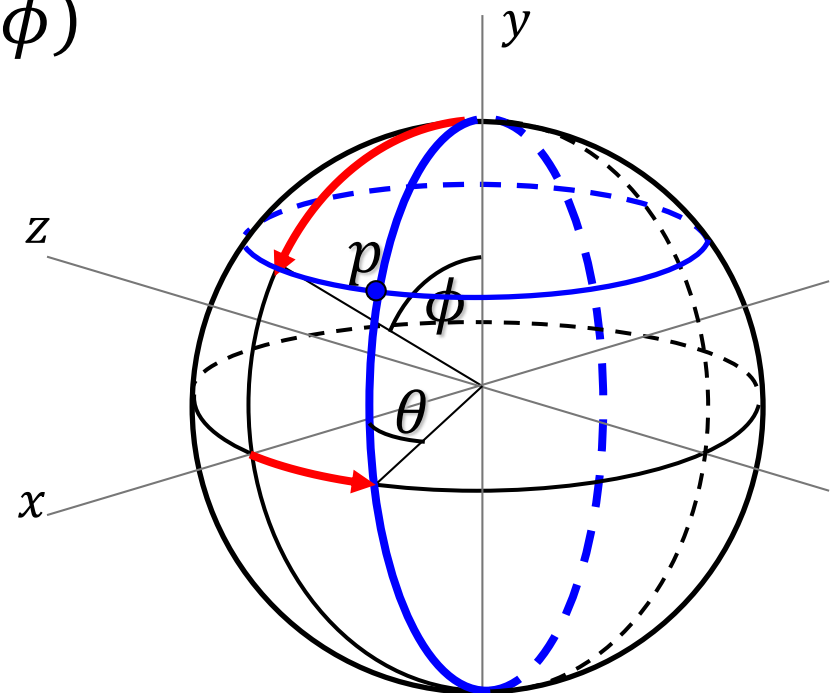


# Spherical Convolution

## Convolution:

Using the Euler angle representation, we know that the rotation taking the North pole to the point  $p = \Phi(\theta, \phi)$  is the rotation:

$$R(\theta, \phi) = R_y(\theta) \cdot R_z(\phi)$$





# Spherical Convolution

## Convolution:

Thus, given

- A spherical function  $g(\theta, \phi)$
- A spherical filter  $f(\theta, \phi)$  that is rotationally-symmetric about the  $y$ -axis

The convolution of  $g$  with  $f$  at  $p = \Phi(\theta, \phi)$  can be expressed by rotating  $f$  so the North pole gets mapped to  $p$  and computing the inner product:

$$(f * g)(\theta, \phi) = \langle \rho_{R(\theta, \phi)}(f), \bar{g} \rangle$$



# Spherical Convolution

## Convolution:

Expressing the spherical functions  $f$  and  $g$  in terms of the spherical harmonic basis, we get:

$$f(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{f}}_{lm} Y_l^m(\theta, \phi)$$
$$g(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{g}}_{lm} Y_l^m(\theta, \phi)$$





# Spherical Convolution

## Convolution:

Recall that the spherical harmonics can be expressed as a complex exponential in  $\theta$  times a “polynomial” in  $\cos \phi$ :

$$Y_l^m(\theta, \phi) = P_l^m(\cos \phi) \cdot e^{im\theta}$$

So a rotation by  $\alpha$  degrees about the  $y$ -axis acts on the  $(l, m)$ -th spherical harmonic by:

$$\rho_{R_y(\alpha)}(Y_l^m) = e^{-im\alpha} \cdot Y_l^m$$



# Spherical Convolution

## Convolution:

If the filter  $f$  is rotationally symmetric about the  $y$ -axis, any rotation about the  $y$ -axis must not change  $f$ . That is, for all  $\alpha$  we must have:

$$\rho_{R_y(\alpha)}(f) = f$$

Or in terms of the spherical harmonics:

$$\sum_l \sum_{m=-l}^l \hat{\mathbf{f}}_{lm} Y_l^m(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{f}}_{lm} \cdot e^{-im\alpha} \cdot Y_l^m(\theta, \phi)$$
$$\Downarrow$$
$$\hat{\mathbf{f}}_{lm} = \hat{\mathbf{f}}_{lm} e^{-im\alpha}$$



# Spherical Convolution

Convolution:

$$\hat{\mathbf{f}}_{lm} = \hat{\mathbf{f}}_{lm} e^{-im\alpha}$$

For this to be true, either:

- $e^{-im\alpha} = 1$  for all  $\alpha \Rightarrow m = 0$ , or
- $\hat{\mathbf{f}}_{lm} = 0$

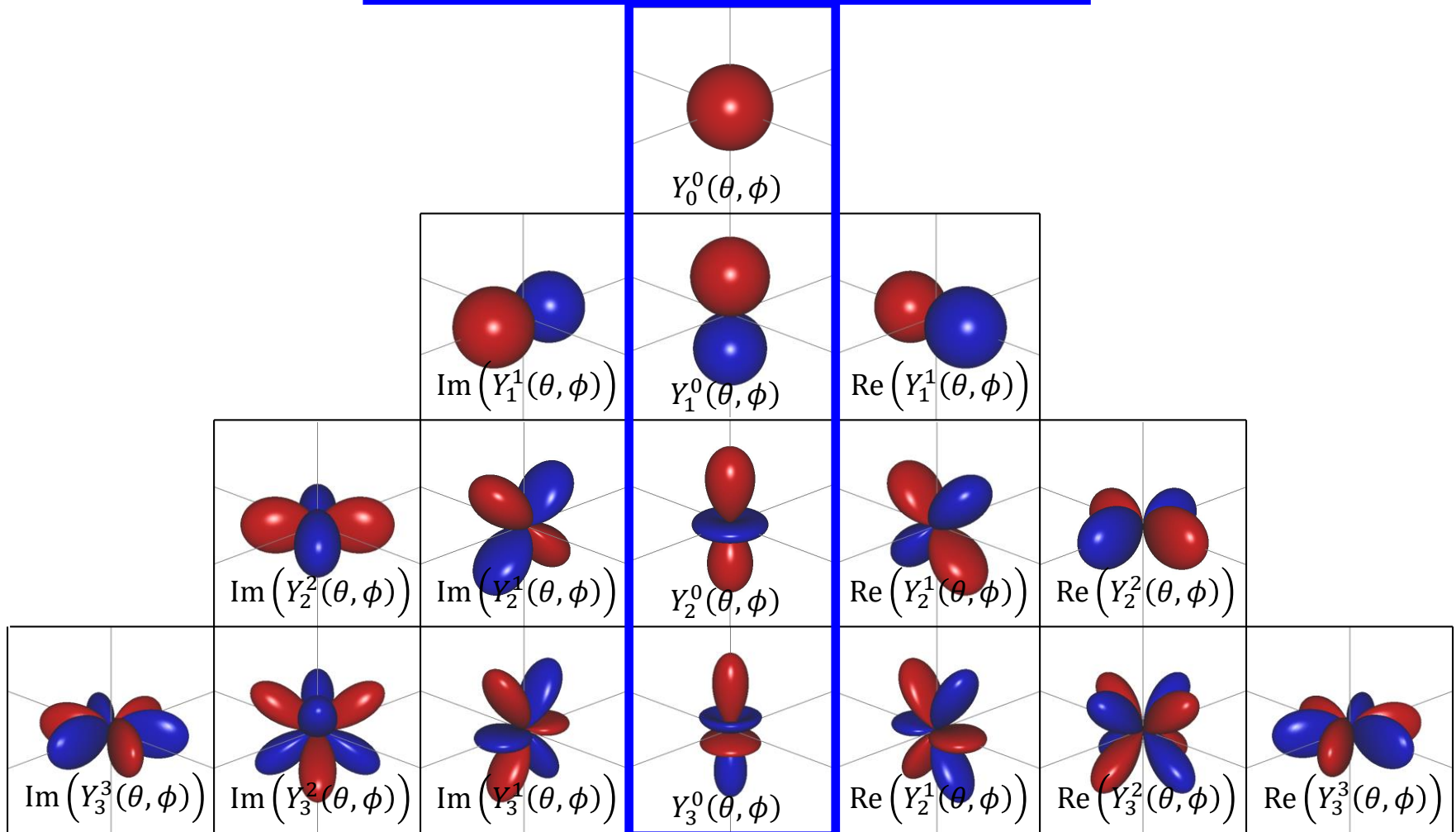
Thus, in terms of the spherical harmonics, we get:

$$f(\theta, \phi) = \sum_l \hat{\mathbf{f}}_{l0} Y_l^0(\theta, \phi)$$



# Spherical Convolution

$$f(\theta, \phi) = \sum_l \hat{f}_{l0} Y_l^0(\theta, \phi)$$





# Spherical Convolution

## Convolution:

Thus, the expression for the functions in terms of their spherical harmonic decomposition becomes:

$$f(\theta, \phi) = \sum_{l'} \hat{\mathbf{f}}_{l'0} Y_{l'}^0(\theta, \phi)$$

$$g(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{g}}_{lm} Y_l^m(\theta, \phi)$$

and we get an expression for the convolution:

$$(f * g)(\theta, \phi) = \left\langle \rho_R(\theta, \phi) \left( \sum_{l'} \hat{\mathbf{f}}_{l'0} Y_{l'}^0 \right), \sum_l \sum_{m=-l}^l \overline{\hat{\mathbf{g}}_{lm} Y_l^m} \right\rangle$$



# Spherical Convolution

Convolution:

$$(f * g)(\theta, \phi) = \left\langle \rho_{R(\theta, \phi)} \left( \sum_{l'} \hat{\mathbf{f}}_{l'0} Y_{l'}^0 \right), \sum_l \sum_{m=-l}^l \overline{\hat{\mathbf{g}}_{lm} Y_l^m} \right\rangle$$

By leveraging the conjugate-linearity of the inner product and using the fact that the transformation  $\rho_R$  is linear, we get:

$$(f * g)(\theta, \phi) = \sum_{l, l'} \sum_{m=-l}^l \hat{\mathbf{f}}_{l'0} \cdot \hat{\mathbf{g}}_{lm} \langle \rho_{R(\theta, \phi)}(Y_{l'}^0), \overline{Y_l^m} \rangle$$



# Spherical Convolution

## Convolution:

Additionally, we know that:

- A rotation of an  $l$ -th frequency function will still be an  $l$ -th frequency function
- The space of  $l$ -th frequency functions is orthogonal to the space of  $l'$ -th frequency functions (if  $l \neq l'$ )

Thus, for all  $l \neq l'$ , we have:

$$\left\langle \rho_R \left( Y_{l'}^{m'} \right), Y_l^m \right\rangle = 0$$



# Spherical Convolution

## Convolution:

This lets us simplify the expression for the convolution:

$$(f * g)(\theta, \phi) = \sum_{l, l'} \sum_{m=-l}^l \hat{\mathbf{f}}_{l'0} \cdot \hat{\mathbf{g}}_{lm} \langle \rho_{R(\theta, \phi)}(Y_{l'}^0), \overline{Y_l^m} \rangle$$
$$\Downarrow$$
$$(f * g)(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{f}}_{l0} \cdot \hat{\mathbf{g}}_{lm} \langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle$$





# Spherical Convolution

Convolution:

$$(f * g)(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{f}}_{l0} \cdot \hat{\mathbf{g}}_{lm} \langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle$$

To compute the convolution, we need to be able to evaluate the inner product:

$$\langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle$$



# Spherical Convolution

## Convolution:

What is the meaning of the function:

$$\langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle$$

This is a function on the sphere whose value at the point  $p = \Phi(\theta, \phi)$  is the dot-product of  $Y_l^m$  with the rotation of  $Y_l^0$ , where the rotation takes the North pole to  $p$ .

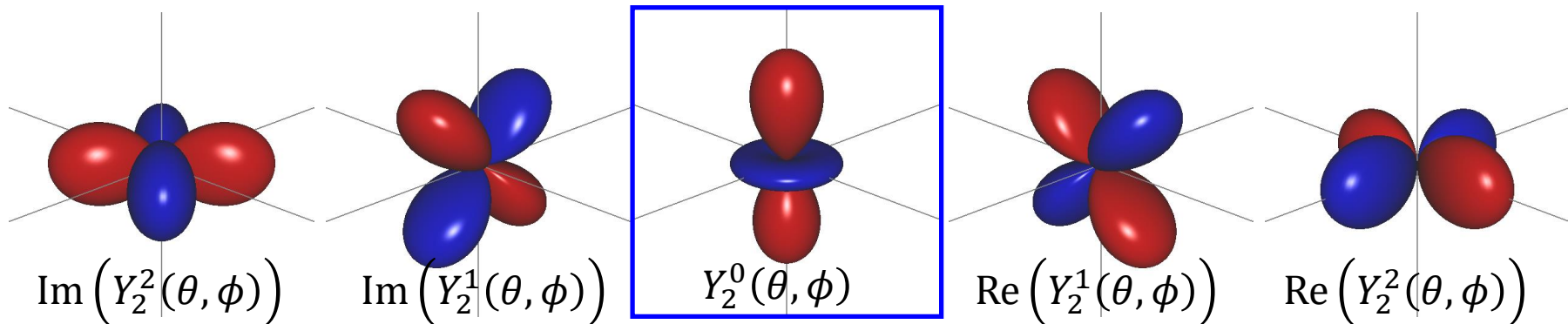


# Spherical Convolution

## Convolution:

We would like to show that this function acts very simply on the spherical harmonics:

$$\langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle = \lambda_l \cdot Y_l^m(\theta, \phi)$$





# Spherical Convolution

## Convolution:

Let's consider the operator  $C_l$  that maps spherical functions to spherical functions, defined by:

$$(C_l(g))(\theta, \phi) = \langle \rho_{R(\theta, \phi)}(Y_l^0), \bar{g} \rangle$$

As before, it turns out this map is a symmetric linear operator on the space of functions.

Thus, there exists an orthonormal basis with respect to which  $C_l$  is diagonal.



# Spherical Convolution

## Convolution:

Let's consider the operator  $C_l$  that maps spherical functions to spherical functions, defined by:

$$(C_l(g))(\theta, \phi) = \langle \rho_{R(\theta, \phi)}(Y_l^0), \bar{g} \rangle$$

This operator also has the property that it commutes with rotations:

- Rotating a spherical function and then convolving with  $Y_l^0$  is the same as first convolving with  $Y_l^0$  and then rotating.

# Spherical Convolution



## Convolution:

So, as with the Laplacian, we have a case in which we are given a symmetric operator which commutes with rotations.



# Spherical Convolution

$L$ : a symmetric operator

$R \in SO(3)$ : a rotation

$V_\lambda$ : the space of e. functions of  $L$  with e.value  $\lambda$

$$R(L(f)) = L(R(f))$$

$$\Downarrow$$

$$\lambda \cdot R(f) = L(R(f)) \quad \forall f \in V_\lambda$$

$$\Downarrow$$

$$R(f) \in V_\lambda$$



# Spherical Convolution

## Convolution:

Thus, the subspace of  $l'$ -th frequency functions is a space of functions that are eigenvectors of  $C_l$ , all with the same eigenvalue:

$$C_l(Y_{l'}^m) = \lambda_{ll'} \cdot Y_{l'}^m$$

Thus we have:

$$C_l(Y_{l'}^m) = Y_{l'}^m \cdot \begin{cases} 0 & \text{if } l \neq l' \\ \lambda_l & \text{otherwise} \end{cases}$$





# Spherical Convolution

## Convolution:

Putting this all together, we get:

$$\langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle = \lambda_l \cdot Y_l^m(\theta, \phi)$$

Thus, the equation for the convolution becomes:

$$(f * g)(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{f}}_{l0} \cdot \hat{\mathbf{g}}_{lm} \cdot \langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle$$

$\Downarrow$

$$(f * g)(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{f}}_{l0} \cdot \hat{\mathbf{g}}_{lm} \cdot \lambda_l \cdot Y_l^m(\theta, \phi)$$



# Spherical Convolution

Convolution:

$$(f * g)(\theta, \phi) = \sum_l \sum_{m=-l}^l \hat{\mathbf{f}}_{l0} \cdot \hat{\mathbf{g}}_{lm} \cdot \lambda_l \cdot Y_l^m(\theta, \phi)$$

Thus, the convolution of  $f$  with  $g$  can be obtained by multiplying the  $(l, m)$ -th spherical harmonic coefficients of  $g$  by  $\lambda_l \cdot \hat{\mathbf{f}}_{l0}$ .

As in the case of functions on a circle, this means that convolution in the spatial domain amounts to multiplication in the frequency domain.



# Spherical Convolution

## Convolution:

In order to be able to use the convolution theorem for spherical functions, we need to know what the eigenvalues  $\lambda_l$  are.

It turns out that these are:

$$\lambda_l = \sqrt{\frac{4\pi}{2l+1}}$$

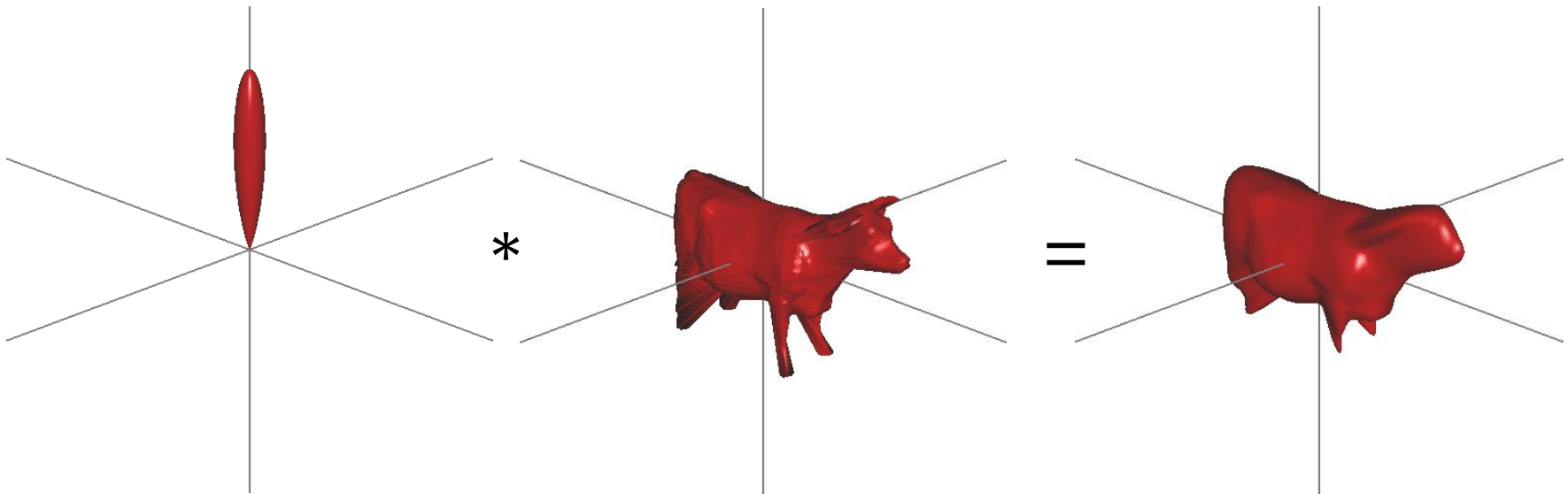


# Spherical Convolution

Convolution:

Which gives us the equation:

$$\langle f * g, Y_l^m \rangle = \sqrt{\frac{4\pi}{2l+1}} \cdot \hat{\mathbf{f}}_{l0} \cdot \hat{\mathbf{g}}_{lm}$$



# Outline

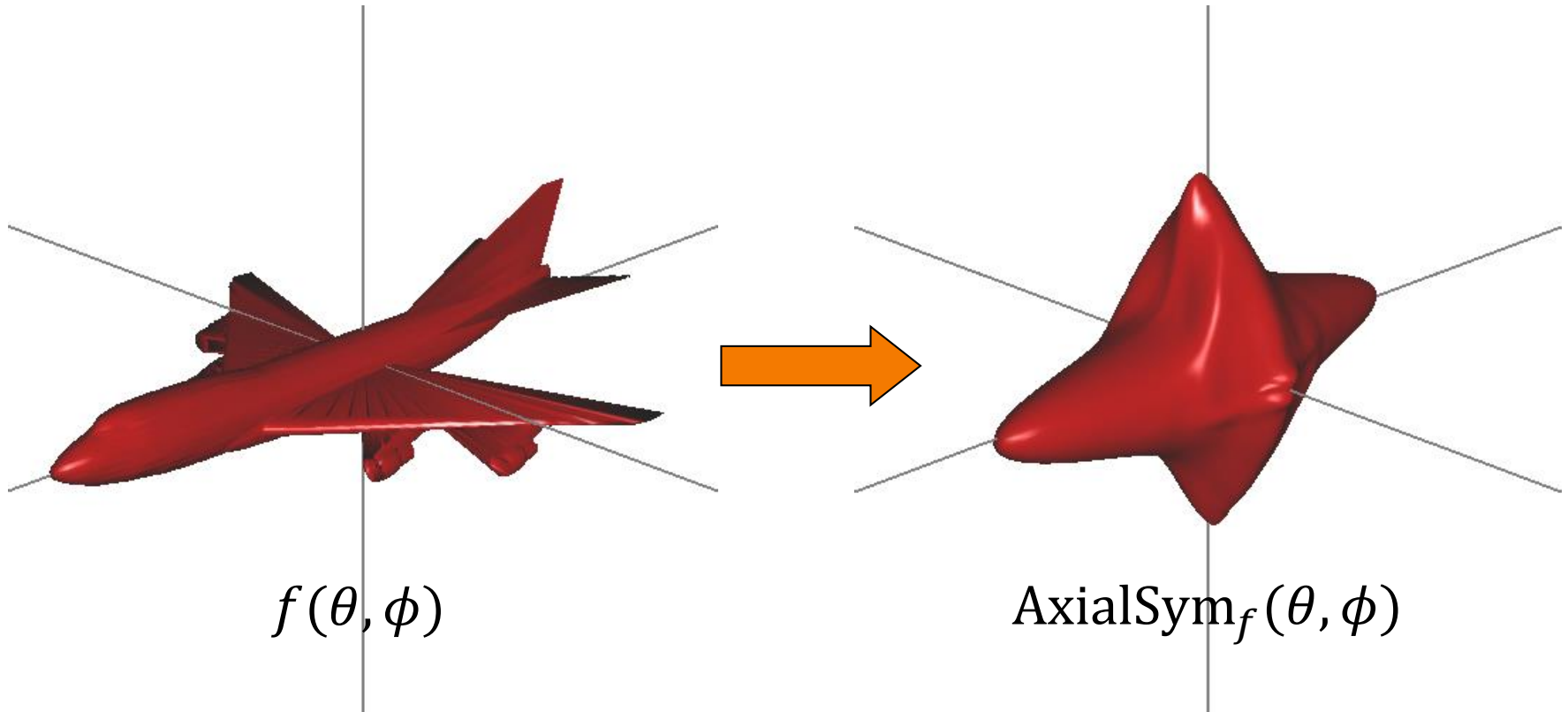
- Math Review
- Spherical Convolution
- Axial Symmetry Detection





# Axial Symmetry Detection

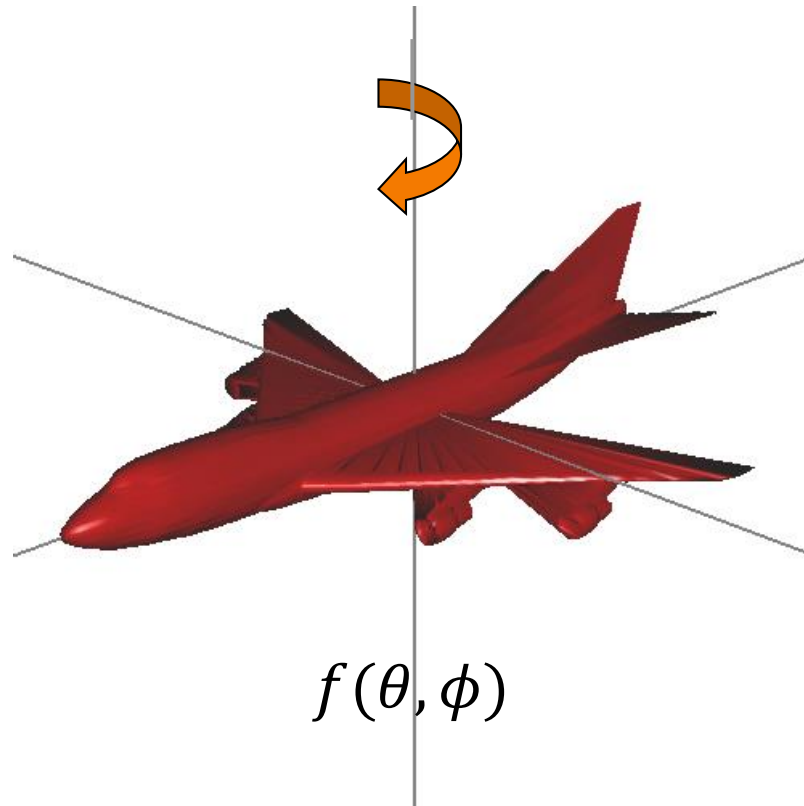
Given a spherical function  $f$ , we would like to compute the measure of the axial symmetry of  $f$  with respect to every axis through the origin.





# Axial Symmetry Detection

What is the measure of the axial symmetry of  $f$  about the  $y$ -axis?

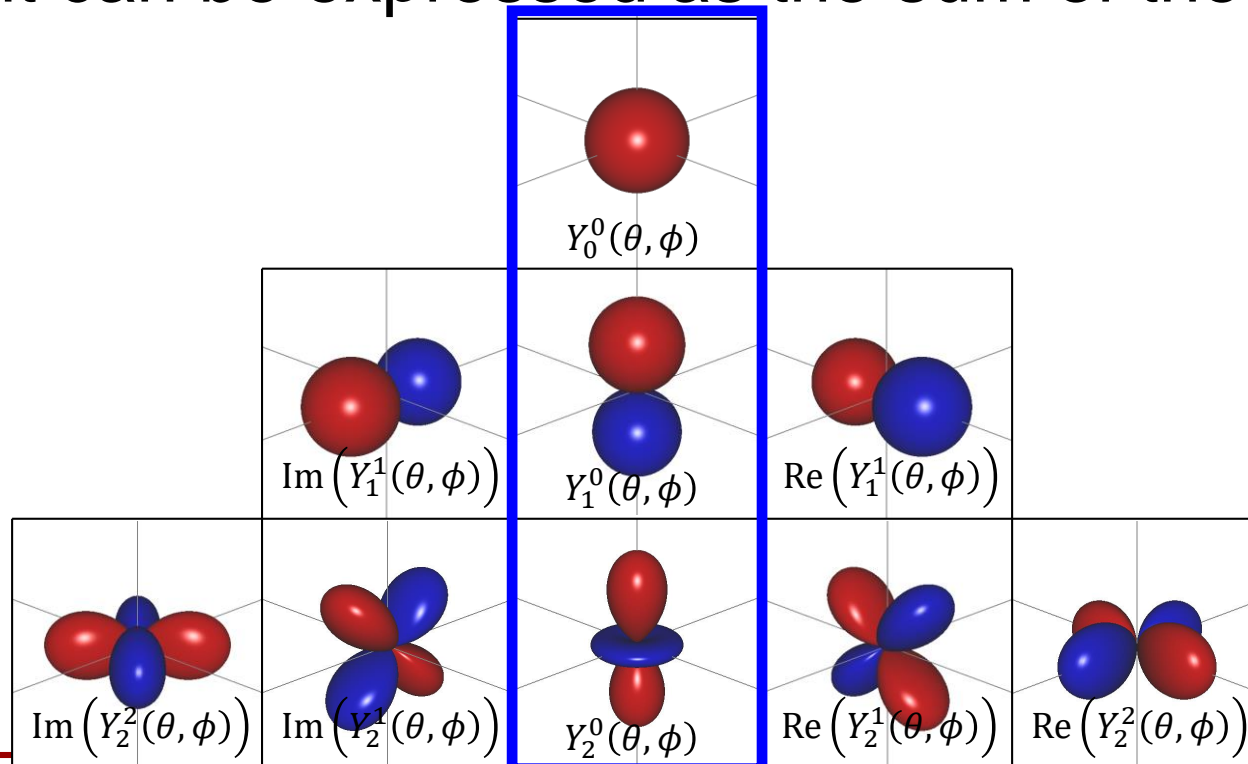




# Axial Symmetry Detection

What is the measure of the axial symmetry of  $f$  about the  $y$ -axis?

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We also know that for  $m \neq 0$ :

$$\langle Y_l^0, Y_l^m \rangle = 0$$

So the projection onto the space of functions that are axially symmetric about the  $y$ -axis is:

$$\pi_y \left( \sum_l \sum_{m=-l}^k \hat{\mathbf{f}}_{lm} Y_l^m \right) = \sum_l \hat{\mathbf{f}}_{l0} Y_l^0$$



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Thus, the measure of the axial symmetry of  $f$  about the  $y$ -axis is defined as:

$$\begin{aligned} \text{YAxialSym}^2(f) &= \left\| \sum_l \hat{\mathbf{f}}_{l0} Y_l^0 \right\|^2 \\ &= \sum_l \|\hat{\mathbf{f}}_{l0}\|^2 \end{aligned}$$

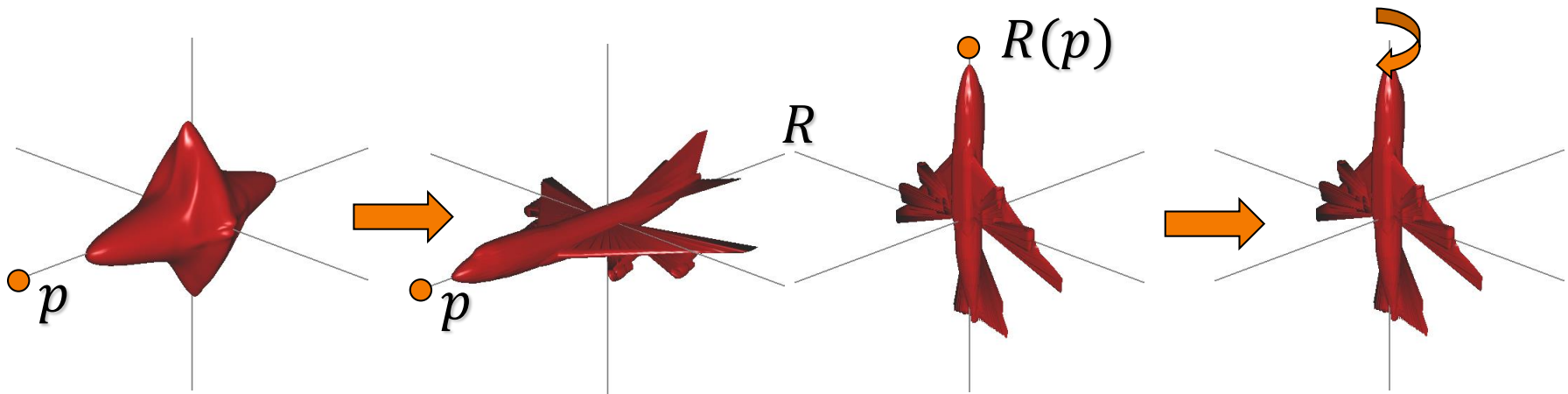


# Axial Symmetry Detection

More generally, we would like to compute the measure of the axial symmetry of  $f$  with respect to any axis.

To compute the symmetry measure about the line through  $p = \Phi(\theta, \phi)$  we:

- Rotate so that  $p$  goes to the North pole, and
- Compute the symmetry measure about the  $y$ -axis.



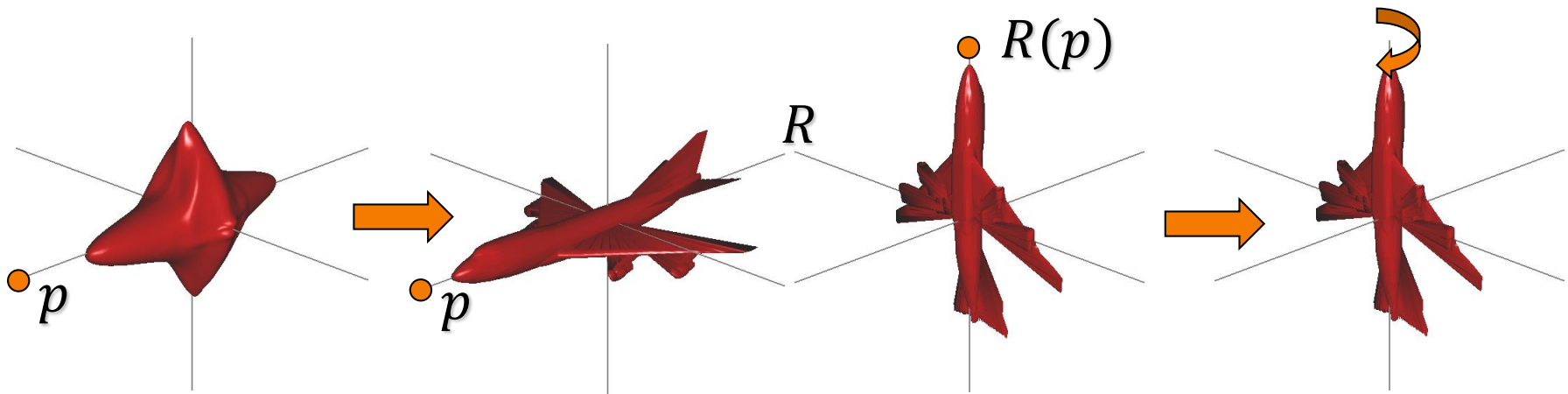


# Axial Symmetry Detection

More generally, we would like to be able to compute the measure of the axial symmetry of  $f$  with

To do this, we are interested in the line through the North pole to  $p$ , the rotation we are interested in is the inverse,  $R^{-1}(\theta, \phi)$ .

- Rotate so that  $p$  goes to the North pole, and
- Compute the symmetry measure about the  $y$ -axis.





# Axial Symmetry Detection

Using the fact that the spherical harmonics form an orthonormal basis, we know that the  $(l, m)$ -th harmonic coefficient of  $f$  is defined by:

$$\hat{\mathbf{f}}_{lm} = \langle f, Y_l^m \rangle$$

Thus, to compute the measure of axial symmetry about the axis through  $p$  we need to compute:

$$\text{AxialSym}_f^2(\theta, \phi) = \sum_l \left\| \left\langle \rho_{R^{-1}(\theta, \phi)}(f), Y_l^0 \right\rangle \right\|^2$$



# Axial Symmetry Detection

$$\text{AxialSym}_f^2(\theta, \phi) = \sum_l \left\| \left\langle \rho_{R^{-1}(\theta, \phi)}(f), Y_l^0 \right\rangle \right\|^2$$

Using the facts that  $\rho$  is a unitary representation and that the zonal harmonics are real-valued, we can re-write this equation as:

$$\begin{aligned} \text{AxialSym}_f^2(\theta, \phi) &= \sum_l \left\| \left\langle f, \rho_{R(\theta, \phi)}(Y_l^0) \right\rangle \right\|^2 \\ &= \sum_l \left\| \left\langle \rho_{R(\theta, \phi)}(Y_l^0), \bar{f} \right\rangle \right\|^2 \end{aligned}$$



# Axial Symmetry Detection

$$\text{AxialSym}_f^2(\theta, \phi) = \sum_l \left\| \langle \rho_{R(\theta, \phi)}(Y_l^0), \bar{f} \rangle \right\|^2$$

Expressing  $f$  in terms of its spherical harmonic decomposition, we get:

$$\text{AxialSym}_f^2(\theta, \phi) = \sum_l \left\| \sum_{m=-l}^l \hat{\mathbf{f}}_{lm} \langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle \right\|^2$$



# Axial Symmetry Detection

$$\text{AxialSym}_f^2(\theta, \phi) = \sum_l \left\| \sum_{m=-l}^l \hat{\mathbf{f}}_{lm} \langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle \right\|^2$$

Applying the identity:

$$\langle \rho_{R(\theta, \phi)}(Y_l^0), \overline{Y_l^m} \rangle = \sqrt{\frac{4\pi}{2l+1}} Y_l^m(\theta, \phi)$$

we get an expression for the symmetry measure:

$$\text{AxialSym}_f^2(\theta, \phi) = \sum_l \frac{4\pi}{2l+1} \left( \left\| \sum_{m=-l}^l \hat{\mathbf{f}}_{lm} Y_l^m(\theta, \phi) \right\|^2 \right)$$





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$$\text{AxialSym}_f^2(\theta, \phi) = \sum_l \frac{4\pi}{2l+1} \left( \left\| \sum_{m=-l}^l \hat{\mathbf{f}}_{lm} Y_l^m(\theta, \phi) \right\|^2 \right)$$

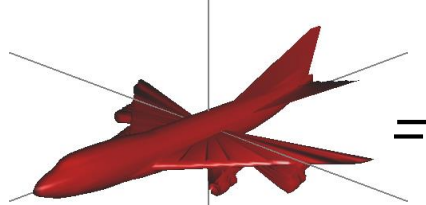
Thus, the measure of axial symmetry can be computed by taking the weighted sum of the squares of the frequency components of  $f$ .



# Axial Symmetry Detection

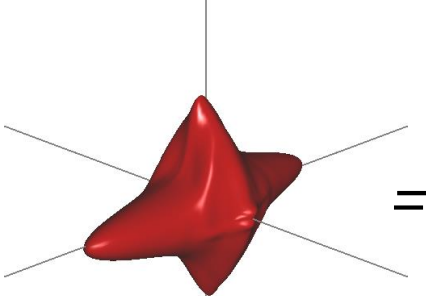
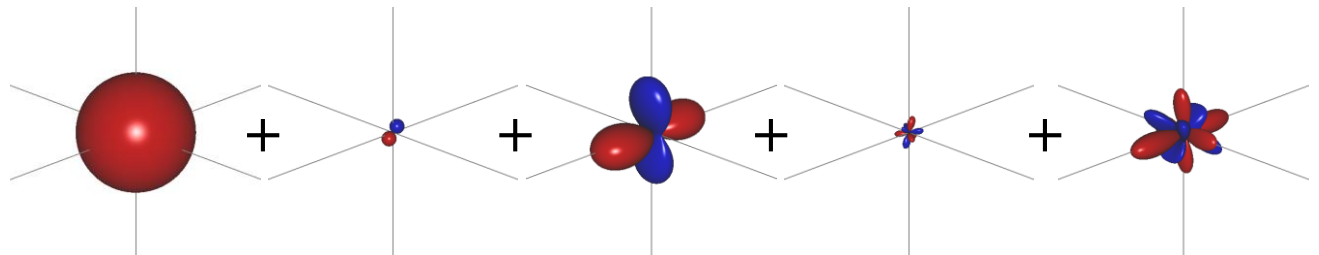
$$\text{AxialSym}_f^2(\theta, \phi) = \sum_l \frac{4\pi}{2l+1} \left( \left\| \sum_{m=-l}^l \hat{\mathbf{f}}_{lm} Y_l^m(\theta, \phi) \right\|^2 \right)$$

Initial Function

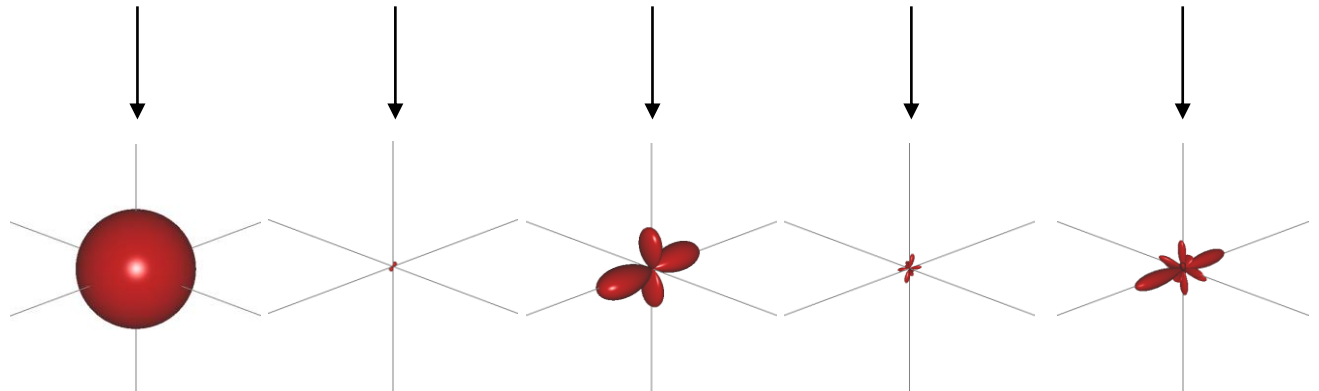


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Frequency Decomposition



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Axial Symmetry  
Descriptor

Weighted Square Norms