

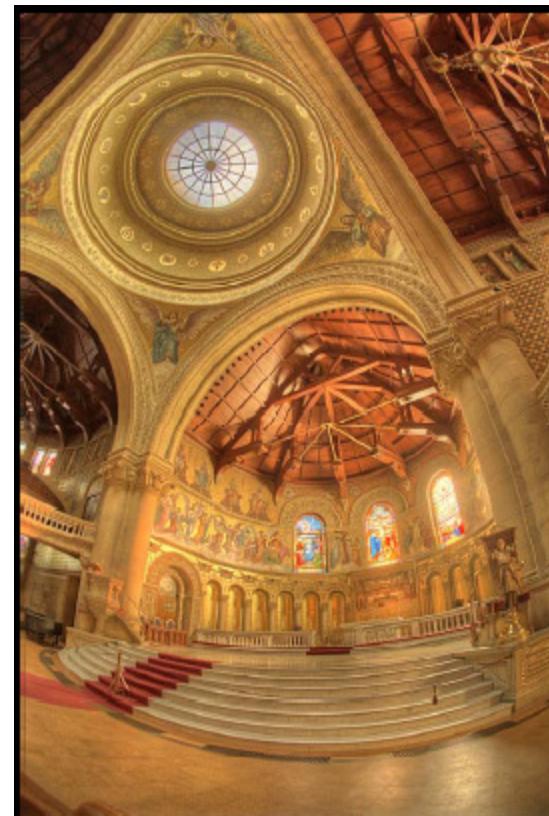
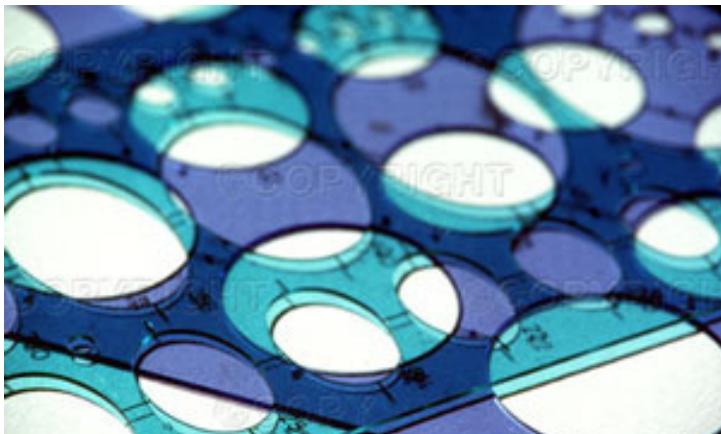
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# Ellipse Fitting

COMP 4900D  
Winter 2006

# Ellipses in Images

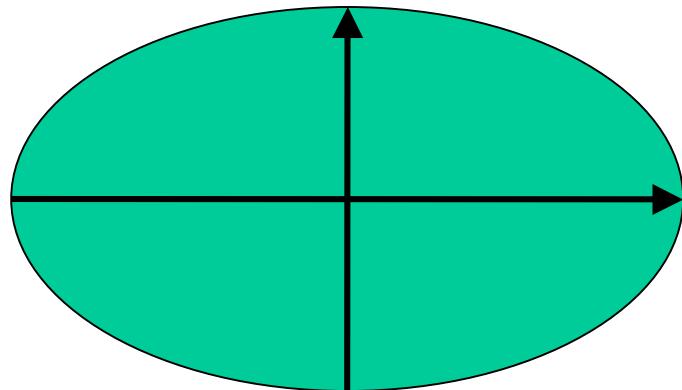
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Perspective projection of circles form ellipses in the images.

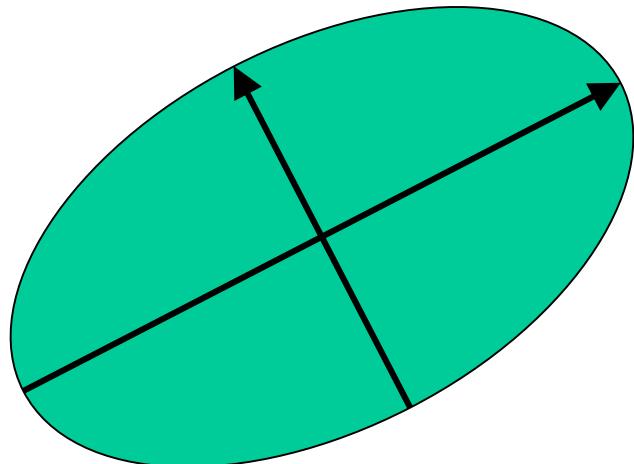
# Equations of Ellipse

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$$\frac{x^2}{r_1^2} + \frac{y^2}{r_2^2} = 1$$

$$ax^2 + bxy + cy^2 + dx + ey + f = 0$$



Let  $\mathbf{x} = [x^2, xy, y^2, x, y, 1]^T$

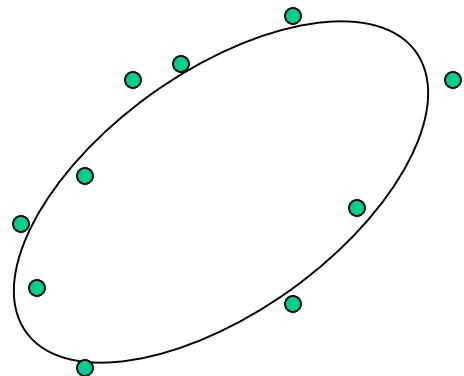
$$\mathbf{a} = [a, b, c, d, e, f]^T$$

Then  $\mathbf{x}^T \mathbf{a} = 0$

# Ellipse Fitting: Problem Statement

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Given a set of  $N$  image points  $\mathbf{p}_i = [x_i, y_i]^T$   
find the parameter vector  $\mathbf{a}_0$  such that the ellipse



$$f(\mathbf{p}, \mathbf{a}) = \mathbf{x}^T \mathbf{a} = 0$$

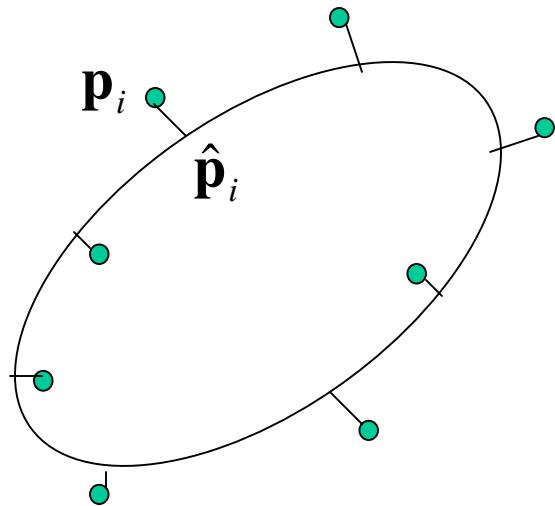
fits  $\mathbf{p}_i$  best in the least square sense:

$$\min_{\mathbf{a}} \sum_{i=1}^N [D(\mathbf{p}_i, \mathbf{a})]^2$$

Where  $D(\mathbf{p}_i, \mathbf{a})$  is the distance from  $\mathbf{p}_i$  to the ellipse.

# Euclidean Distance Fit

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$$D(\mathbf{p}_i, \mathbf{a}) = \|\hat{\mathbf{p}}_i - \mathbf{p}_i\|$$

$\hat{\mathbf{p}}_i$  is the point on the ellipse that is nearest to  $\mathbf{p}_i$

$$f(\hat{\mathbf{p}}_i, \mathbf{a}) = 0$$

$\hat{\mathbf{p}}_i - \mathbf{p}_i$  is normal to the ellipse at  $\hat{\mathbf{p}}_i$

# Compute Distance Function

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Computing the distance function is a constrained optimization problem:

$$\min_{\hat{\mathbf{p}}_i} \|\hat{\mathbf{p}}_i - \mathbf{p}_i\|^2 \quad \text{subject to} \quad f(\hat{\mathbf{p}}_i, \mathbf{a}) = 0$$

Using **Lagrange multiplier**, define:

$$L(x, y, \lambda) = \|\hat{\mathbf{p}}_i - \mathbf{p}_i\|^2 - 2\lambda f(\hat{\mathbf{p}}_i, \mathbf{a})$$

where  $\hat{\mathbf{p}}_i = [x, y]^T$

Then the problem becomes:  $\min_{\hat{\mathbf{p}}_i} L(x, y, \lambda)$

Set  $\frac{\partial L}{\partial x} = \frac{\partial L}{\partial y} = 0$  we have  $\hat{\mathbf{p}}_i - \mathbf{p}_i = \lambda \nabla f(\hat{\mathbf{p}}_i, \mathbf{a})$

# Two Approximations

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1. First-order approximation at  $\mathbf{p}_i$

$$f(\hat{\mathbf{p}}_i, \mathbf{a}) \approx f(\mathbf{p}_i, \mathbf{a}) + (\hat{\mathbf{p}}_i - \mathbf{p}_i)^T \nabla f(\mathbf{p}_i, \mathbf{a}) = 0$$

2. Assume  $\mathbf{p}_i$  is close to  $\hat{\mathbf{p}}_i$ , then

$$\nabla f(\hat{\mathbf{p}}_i, \mathbf{a}) \approx \nabla f(\mathbf{p}_i, \mathbf{a})$$

# Approximate Distance Function

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$$f(\hat{\mathbf{p}}_i, \mathbf{a}) \approx f(\mathbf{p}_i, \mathbf{a}) + (\hat{\mathbf{p}}_i - \mathbf{p}_i)^T \nabla f(\mathbf{p}_i, \mathbf{a}) = 0$$

$$\hat{\mathbf{p}}_i - \mathbf{p}_i = \lambda \nabla f(\hat{\mathbf{p}}_i, \mathbf{a}) \approx \lambda \nabla f(\mathbf{p}_i, \mathbf{a})$$

Solve for  $\lambda$

$$\lambda = -\frac{f(\mathbf{p}_i, \mathbf{a})}{\|\nabla f(\mathbf{p}_i, \mathbf{a})\|^2}$$

Substitute back

$$\hat{\mathbf{p}}_i - \mathbf{p}_i = -\frac{f(\mathbf{p}_i, \mathbf{a}) \nabla f(\mathbf{p}_i, \mathbf{a})}{\|\nabla f(\mathbf{p}_i, \mathbf{a})\|^2}$$

$$D(\mathbf{p}_i, \mathbf{a}) = \|\hat{\mathbf{p}}_i - \mathbf{p}_i\| = \frac{|f(\mathbf{p}_i, \mathbf{a})|}{\|\nabla f(\mathbf{p}_i, \mathbf{a})\|}$$

# Ellipse Fitting with Euclidean Distance

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Given a set of  $N$  image points  $\mathbf{p}_i = [x_i, y_i]^T$   
find the parameter vector  $\mathbf{a}_0$  such that

$$\min_{\mathbf{a}} \sum_{i=1}^N \frac{|f(\mathbf{p}_i, \mathbf{a})|^2}{\|\nabla f(\mathbf{p}_i, \mathbf{a})\|^2}$$

This problem can be solved by using a numerical nonlinear optimization system.