

1 Tylor expansion of Images

The promblem of computing mage derivatives arises from a least square error minimization of the energy function (1-d case)

$$E = \sum_{i=-n}^n (I(n) - \tilde{I}(n))^2 \quad (1)$$

Where \tilde{I} is the Taylor series approximation of I .

Recall that the Taylor series expansion of a function $f(x)$ about a point a is given by

$$\sum_{n=0}^{\infty} \frac{f^{(n)}(a)}{n!} \cdot (x - a)^n$$

1.1 The 2-d case

Consider the second order Tylor series expansion about a point (x_0, y_0) in a 2-d image

$$\tilde{I}(x, y) = I + I_x \cdot (x - x_0) + I_y \cdot (y - y_0) \quad (2)$$

$$+ \frac{1}{2} [I_{xx} \cdot (x - x_0)^2 + 2 \cdot I_{xy} \cdot (x - x_0)(y - y_0) + I_{yy} \cdot (y - y_0)^2] \quad (3)$$

Where $I_x = I_x(x_0, y_0)$ has been used to keep the notation short.

This expansion can be expressed in vector-matrix form as,

$$\tilde{I}(x, y) = \begin{bmatrix} (x - x_0) & (y - y_0) & 1 \end{bmatrix} \frac{1}{2} \begin{bmatrix} I_{xx} & I_{xy} & I_x \\ I_{xy} & I_{yy} & I_y \\ I_x & I_y & 2I \end{bmatrix} \begin{bmatrix} (x - x_0) \\ (y - y_0) \\ 1 \end{bmatrix}$$

1.2 The 3-d case

In general, the second-order Taylor series expansion, about a d-dimensional point \mathbf{a} , of multivariate functions can be written as,

$$\tilde{f}(\mathbf{x}) = f(\mathbf{a}) + (\mathbf{x} - \mathbf{a})^T \nabla f(\mathbf{a}) + \frac{1}{2!} (\mathbf{x} - \mathbf{a})^T \{ \nabla \nabla f(\mathbf{a}) \} (\mathbf{x} - \mathbf{a})$$

Where, $\nabla f(\mathbf{a})$ is the gradiant of f evaluated at $\mathbf{x} = \mathbf{a}$, and $\nabla \nabla f(\mathbf{a})$ is the Hessian matrix of f evaluated at $\mathbf{x} = \mathbf{a}$,

So, for a 3-d image we obtain

$$\tilde{I}(\mathbf{x}) = I(\mathbf{a}) + (\mathbf{x} - \mathbf{a})^T \begin{bmatrix} I_x(\mathbf{a}) \\ I_y(\mathbf{a}) \\ I_z(\mathbf{a}) \end{bmatrix} + \frac{1}{2!} (\mathbf{x} - \mathbf{a})^T \begin{bmatrix} I_{xx} & I_{xy} & I_{xz} \\ I_{yx} & I_{yy} & I_{yz} \\ I_{zx} & I_{zy} & I_{zz} \end{bmatrix} (\mathbf{x} - \mathbf{a})$$

1.3 Kernel Approximation

Ideally the expansion would agree exactly with intensity values, but since this is not the case, the best we can do can be found with a least squares approximation.

$$E = \sum_{i=-ni}^{ni} \sum_{j=-nj}^{nj} \sum_{k=-nk}^{nk} \left(I(i, j, k) - \tilde{I}(i, j, k) \right)^2$$

1.3.1 Notation and assumptions

Assume $ni = nj = nk$

Assume $\mathbf{a} = \mathbf{0}$

$I(\mathbf{x}) = I(i, j, k)$

$\mathbf{G} = \nabla f(\mathbf{a})$ - The gradient vector

$\mathbf{H} = \nabla \nabla f(\mathbf{a})$ - The hessian matrix

1_{ij} : Single-entry matrix; 1 at (i, j) and zero elsewhere

$$\frac{\partial(\mathbf{x}^T \mathbf{G})}{\partial G_i} = \mathbf{x}^T \frac{\partial(\mathbf{G})}{\partial G_i} = \mathbf{x}^T \mathbf{1}_i$$

$$\frac{\partial(\mathbf{x}^T \mathbf{H} \mathbf{x})}{\partial H_{ij}} = \mathbf{x}^T \frac{\partial(\mathbf{H})}{\partial H_{ij}} \mathbf{x} = \mathbf{x}^T \mathbf{1}_{ij} \mathbf{x}$$

$$E = \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right)^2$$

The previous expression can be minimized to find the best values for $I_0, I_x, I_y, I_{xx}, I_{xy}$ and so on. The minimization is given below:

$$\frac{\partial E}{\partial I_0} = -2 \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_x} = -2 \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n \mathbf{x}^T \mathbf{1}_1 \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_y} = -2 \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n \mathbf{x}^T \mathbf{1}_2 \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_z} = -2 \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n \mathbf{x}^T \mathbf{1}_3 \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_{xx}} = -\frac{2}{2!} \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n (\mathbf{x}^T \mathbf{1}_{(1,1)} \mathbf{x}) \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_{yy}} = -\sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n (\mathbf{x}^T \mathbf{1}_{(2,2)} \mathbf{x}) \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_{zz}} = -\sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n (\mathbf{x}^T \mathbf{1}_{(3,3)} \mathbf{x}) \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_{xy}} = -\sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n (\mathbf{x}^T \mathbf{1}_{(1,2),(2,1)} \mathbf{x}) \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_{xz}} = - \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n (\mathbf{x}^T \mathbf{1}_{(1,3),(3,1)} \mathbf{x}) \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\frac{\partial E}{\partial I_{yz}} = - \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n (\mathbf{x}^T \mathbf{1}_{(2,3),(3,2)} \mathbf{x}) \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

Expanding the previous equations we obtain:

$$\frac{\partial E}{\partial I_0} = -2 \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\sum_{i,j,k} \left(I(i, j, k) - I_0 - (iI_x + jI_y + kI_z) - \frac{1}{2!} (i^2 I_{xx} + j^2 I_{yy} + k^2 I_{zz} + 2ijI_{xy} + 2ikI_{xz} + 2jkI_{yz}) \right)$$

$$\begin{aligned} & \sum_{i,j,k} I(i, j, k) - I_0 \sum_{i,j,k} - (I_x \sum_i i \sum_{j,k} + I_y \sum_j j \sum_{i,k} + I_z \sum_k k \sum_{i,j}) \\ & - \frac{1}{2!} (I_{xx} \sum_i i^2 \sum_{j,k} + I_{yy} \sum_j j^2 \sum_{i,k} + I_{zz} \sum_k k^2 \sum_{i,j} + 2I_{xy} \sum_{i,j} ij \sum_k + 2I_{xz} \sum_{i,k} ik \sum_j + 2I_{yz} \sum_{j,k} jk \sum_i) \end{aligned} \quad (4)$$

$$\frac{\partial E}{\partial I_x} = -2 \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n \mathbf{x}^T \mathbf{1}_1 \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\sum_{i,j,k} i \left(I(i, j, k) - I_0 - (iI_x + jI_y + kI_z) - \frac{1}{2!} (i^2 I_{xx} + j^2 I_{yy} + k^2 I_{zz} + 2ijI_{xy} + 2ikI_{xz} + 2jkI_{yz}) \right)$$

$$\begin{aligned} & \sum_i i \sum_{j,k} I(i, j, k) - I_0 \sum_i i \sum_{j,k} - (I_x \sum_i i^2 \sum_{j,k} + I_y \sum_j j \sum_i i \sum_k + I_z \sum_k k \sum_i i \sum_j) \\ & - \frac{1}{2!} [I_{xx} \sum_i i^3 \sum_{j,k} + I_{yy} \sum_j j^2 \sum_i i \sum_k + I_{zz} \sum_k k^2 \sum_i i \sum_j \\ & + 2I_{xy} \sum_{i,j} i^2 j \sum_k + 2I_{xz} \sum_{i,k} i^2 k \sum_j + 2I_{yz} \sum_{j,k} jk \sum_i i] \end{aligned} \quad (5)$$

$$\frac{\partial E}{\partial I_{xx}} = -\frac{2}{2!} \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n (\mathbf{x}^T \mathbf{1}_{(1,1)} \mathbf{x}) \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\sum_{i,j,k} i^2 \left(I(i, j, k) - I_0 - (iI_x + jI_y + kI_z) - \frac{1}{2!} (i^2 I_{xx} + j^2 I_{yy} + k^2 I_{zz} + 2ijI_{xy} + 2ikI_{xz} + 2jkI_{yz}) \right)$$

$$\begin{aligned} & \sum_i i^2 \sum_{j,k} I(i, j, k) - I_0 \sum_i i^2 \sum_{j,k} - (I_x \sum_i i^3 \sum_{j,k} + I_y \sum_j j \sum_i i^2 \sum_k + I_z \sum_k k \sum_i i^2 \sum_j) \\ & - \frac{1}{2!} [I_{xx} \sum_i i^4 \sum_{j,k} + I_{yy} \sum_j j^2 \sum_i i^2 \sum_k + I_{zz} \sum_k k^2 \sum_i i^2 \sum_j \\ & + 2I_{xy} \sum_{i,j} i^3 j \sum_k + 2I_{xz} \sum_{i,k} i^3 k \sum_j + 2I_{yz} \sum_{j,k} jk \sum_i i^2] \end{aligned} \quad (6)$$

$$\frac{\partial E}{\partial I_{xy}} = - \sum_{i=-n}^n \sum_{j=-n}^n \sum_{k=-n}^n (\mathbf{x}^T \mathbf{1}_{(1,2),(2,1)} \mathbf{x}) \left(I(\mathbf{x}) - I_0 - \mathbf{x}^T \mathbf{G} - \frac{1}{2!} \mathbf{x}^T \mathbf{H} \mathbf{x} \right) = 0$$

$$\sum_{i,j,k} 2ij \left(I(i,j,k) - I_0 - (iI_x + jI_y + kI_z) - \frac{1}{2!}(i^2I_{xx} + j^2I_{yy} + k^2I_{zz} + 2ijI_{xy} + 2ikI_{xz} + 2jkI_{yz}) \right)$$

$$\begin{aligned} & 2 \sum_i i \sum_j j \sum_k k I(i,j,k) - 2I_0 \sum_i i \sum_j j \sum_k k - 2(I_x \sum_i i^2 \sum_j j \sum_k k + I_y \sum_j j^2 \sum_i i \sum_k k + I_z \sum_k k^2 \sum_i i \sum_j j) \\ & - (I_{xx} \sum_i i^3 \sum_j j \sum_k k + I_{yy} \sum_j j^3 \sum_i i \sum_k k + I_{zz} \sum_k k^3 \sum_i i \sum_j j \\ & + 2I_{xy} \sum_i i^2 \sum_j j^2 \sum_k k + 2I_{xz} \sum_i i^2 \sum_k k^2 \sum_j j + 2I_{yz} \sum_j j^2 \sum_k k^2 \sum_i i) \end{aligned} \quad (7)$$

The previous equation indicates that for a second order approximation our Taylor-basis is 9-dimensional
The system of linear equations can be solved using the following matrix:

$$\frac{\delta E}{\delta I_0} \begin{pmatrix} \sum_{ij,k} \\ \sum_i \sum_{j,k} \\ \sum_j \sum_{i,k} \\ \sum_k \sum_{i,j} \\ \sum_i \sum_{j,k}^2 \\ \sum_j \sum_{i,k}^2 \\ \sum_k \sum_{i,j}^2 \\ 2 \sum_i \sum_{j,k}^2 \\ 2 \sum_i \sum_{j,k}^2 \\ 2 \sum_j \sum_{i,k}^2 \end{pmatrix} \begin{pmatrix} I_0 & I_x & I_y & I_z & I_{xx} & I_{yy} & I_{zz} & I_{xy} & I_{xz} & I_{yz} \end{pmatrix} = \begin{pmatrix} \sum_{ij,k} I(i,j,k) \\ \sum_i \sum_{j,k} I(i,j,k) \\ \sum_j \sum_{i,k} I(i,j,k) \\ \sum_k \sum_{i,j} I(i,j,k) \\ \sum_i \sum_{j,k}^2 I(i,j,k) \\ \sum_j \sum_{i,k}^2 I(i,j,k) \\ \sum_k \sum_{i,j}^2 I(i,j,k) \\ 2 \sum_i \sum_{j,k}^2 I(i,j,k) \\ 2 \sum_i \sum_{j,k}^2 I(i,j,k) \\ 2 \sum_j \sum_{i,k}^2 I(i,j,k) \end{pmatrix}$$

Missing! What are the kernels? Well we could run derivatives of gaussians that are already implemented - not the xy. Also, should we go above 2nd derivative - so we get a larger basis to compare Harr Wavelets? or other wavelets?