

Linear Filtering

Convolutions

Flip horizontally and vertically Gaussian Filters:
 $\sum_k G(k) = 1$

Separable Filters

$I \otimes F$
 $= I \otimes (A + B)$
 $= I \otimes (A_1) \otimes (A_2) + I \otimes (B_1) \otimes (B_2)$

Edge Finding

Filters:

$I(x) = G_{x,x}(x, \sigma) \otimes G(x, 3\sigma)$
 $G_{x,x} = \begin{bmatrix} 1 & -1 \end{bmatrix}$
 $I(y) = G_{y,y}(y, \sigma) \otimes (G)$
 $G_{y,y} = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$
 $\theta_{edge} = \tan^{-1}(-\frac{\delta f}{\delta x} \div \frac{\delta f}{\delta y})$

Canny Edge Detection

- 1. Compute image gradient
 $J_x = I \otimes (\frac{\delta}{\delta x} \otimes G)$
- 2. Compute edge gradient magnitude $\nabla J = (J_x, J_y)$
 $\|\nabla J\| = \sqrt[2]{J_x^2 + J_y^2}$

Sampling

Blur then sample

Gaussian Pyramid

Reductions pick weighting function $w_0 = a$
 $w_{-1} = w_1 = 1/4$ $w_{-2} = w_2 = 1/4 - a$ $g_0 = Image$

$g_{L,(i,j)} = \sum_{m=-2}^2 \sum_{n=-2}^2 w(m,n)g_{L-1}(2i+m, 2j+n)$
 $w(m,n) is the 2d filter$
 $g_L = g_{L-1} \otimes w_{downarrow 2}$

Expansion

$g_{L,(i,j)} = 4 \sum_{m=-2}^2 \sum_{n=-2}^2 w(m,n)\hat{g}_{L,n-1}(\frac{i-m}{2}, \frac{j-n}{2})$

Warping

Matricies

Inverse of a Matrix
 $A^{-1} = \frac{1}{det} A^T$
cross product
 $\begin{bmatrix} i & j & k \\ x_1 & x_2 & x_3 \\ y_1 & y_2 & y_3 \end{bmatrix}$
 $= (x_2y_3 - x_3y_2)i + (x_3y_1 - x_1y_3)j + (x_1y_2 - x_2y_1)k$
Translation
 $\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & t_x \\ 0 & 1 & t_y \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$
Scaling
 $\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} a & 0 & 0 \\ 0 & b & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$
Rotation
 $\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} \cos(\theta) & -\sin(\theta) & 0 \\ \sin(\theta) & \cos(\theta) & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$
Skew Shear
 $\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & sh_x & 0 \\ sh_y & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix}$

Affine Transformations Need 3 points Projective Transformations Need 4 points

Triangle Warping to find Homography

ONLY FOR AFFINE TRANSFORMATIONS Warp 3 source points to unit triangle ($T1^{-1}$) then warp unit triangle to destination points (T2)

$T1 = \begin{bmatrix} A-C & B-A & A \\ 0 & 0 & 1 \end{bmatrix}$ $T2 = \begin{bmatrix} A'-C' & B'-A' & A' \\ 0 & 0 & 1 \end{bmatrix}$
 $T1^{-1}T2 = H$

Random Lin alg

A line can be defined as
 $l = x \times x'$ (cross product)
Intersection
 $x = l \times l'$
Line mapping
 $l' = H^{-T}l$

Bilinear Interpolation

$a = i + 1$
 $b = j + 1$
 $f(x, y) = f((1-a), (1-b)) + f(a, (1-b)) + f(a, b) + f((1-a), b)$

Barycentric Coordinates

if $0 \leq a \leq 1$ and $0 \leq a' \leq 1$ then x is inside the triangle

$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} a_x & b_x & c_x \\ a_y & b_y & c_y \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \alpha \\ \beta \\ \gamma \end{bmatrix}$ $\begin{bmatrix} x' \\ y' \\ 1 \end{bmatrix} = \begin{bmatrix} a'_x & b'_x & c'_x \\ a'_y & b'_y & c'_y \\ 1 & 1 & 1 \end{bmatrix} \begin{bmatrix} \alpha' \\ \beta' \\ \gamma' \end{bmatrix}$

TPS

$f(x) = a_1 + a_x(x_im) + a_y(y_im) + \sum_{i=1}^p w_i U(||x_{ctrl_i}, y_{ctrl_i} - (x_im, y_im)||)$
Where p is how many mean control points there are x and y are pixel position in image x_{ctrl}, y_{ctrl} are mean control points
 $U = -r^2 \log 10r^2$ where r is the l2 norm between control points at i and the pixel w is the weighting

Blending

Alpha Blending

$I_{blend} = \alpha I_{for} + (1 - \alpha)I_{back}$
Where α is an alpha mask and multiplication is component wise
For multiple blending
premultiply alpha to all color channels use first equation

Two Band Blending

- 1. make a binary mask apply it to foreground high pass image and 1 - mask to high pass background image
- 2. make linear mask by applying gauss filter to binary mask
- 3. apply linear filter to foreground and 1-linear filter to back ground
- 4. add all of the things

Seam Carving

Energy map
 $e(Image) = |\frac{\delta}{\delta x} I| + |\frac{\delta}{\delta y} I|$
TA's just took the gradient of image and get X gradient matrix and Y gradient matrix and added them together
optimal seam:
 $s* = min_s \sum_{i=1}^n e(I(s_i))$
Cumulative Map for vertical seam switch indices for horizontal seam
 $Map(i, j) = e(i, j) + min(M(i-1, j-1), M(i-1, j), M(i-1, j+1))$
Then start from bottom of Map and select the lowest value pixel

Harris Corners

$M = \sum_{x,y} \begin{bmatrix} I_x^2 & I_x I_y \\ I_x I_y & I_y^2 \end{bmatrix}$

Eigen values of M λ_1, λ_2
Corner Response
 $R = det M \ Trace M \ det M = \lambda_1 * \lambda_2$
Trace M = $\lambda_1 + \lambda_2$
edge if $\lambda_1 or \lambda_2$ are large
corner if both are large

RANSAC

$N = \frac{\log(1-p)}{\log(1-(1-e)^s)}$

p = desired probability that random sample is free from outliers
 $e = 1 - p$ s = number of points in sample N = number of loops in RANSAC (total samples)

Early Termination

$T = 1 - e * totalnumberofdatapoints$