

Image processing by deformation

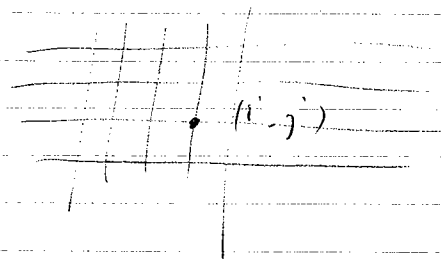
Public lecture

- First talk: "how to view Gaussian filter as a deformation process";
- 2) Introducing level set method.

Example 1: Gaussian diffusion: $U(x, y, t)$

$$(1) \begin{cases} U_t = \Delta U = U_{xx} + U_{yy} & \Omega \times (0, \infty) \\ U(x, y, 0) = I(x, y) & \Omega \\ \frac{\partial U}{\partial \nu} = \nabla U \cdot \nu = 0 & \partial \Omega \end{cases}$$

Grid of an image



m moments

$$U^m(i, j) := U(i, j, m)$$

$$\begin{aligned} \text{Then } U_{xx}(i, j, n) &= \frac{U^n(i+1, j) + U^n(i-1, j) - 2U^n(i, j)}{\Delta x^2} \\ &= \frac{U_{i+1, j}^n + U_{i-1, j}^n - 2U_{i, j}^n}{\Delta x^2} \end{aligned}$$

$$U_{yy}(i, j, n) = \frac{U_{i, j+1}^n + U_{i, j-1}^n - 2U_{i, j}^n}{\Delta y^2}$$

$$U_t(i, j, n) = \frac{U(i, j, n+1) - U(i, j, n)}{\Delta t} = \frac{U_{i, j}^{n+1} - U_{i, j}^n}{\Delta t}$$

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$$\Rightarrow U_{i,j}^{n+1} = \frac{\Delta t}{\Delta x^2} \left[U_{i+1,j}^n + U_{i-1,j}^n + U_{i,j+1}^n + U_{i,j-1}^n - 4U_{i,j}^n + U_{i,j}^n \right]$$

Or. $U_{i,j}^{n+1} = U_{i,j}^n + \text{step} \cdot \Delta U.$

Why it works? When ^{does} the equation comes from? (12:10:00)

Consider functional

$$F(u) = \int_{\Omega} |\nabla u|^2 dx dy = \int_{\Omega} (u_x^2 + u_y^2) dx dy.$$

Find a path to decrease the value of $F(u)$:

Recall: from calculus course:

$$Z = f(x, y): \mathbb{R}^2 \rightarrow \mathbb{R}, \quad \nu = (\nu_1, \nu_2) \text{ unit vector}$$

Change of f along ν direction (Directional derivative)

$$\partial_{\nu} f := \nabla f \cdot \nu. \quad \text{so if } \nu = \begin{cases} \frac{\nabla f}{|\nabla f|} \\ -\frac{\nabla f}{|\nabla f|} \end{cases} \text{ negative}$$

f decreasing or increasing faster.

A bit complicated for functional:

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Public lecture

For a given $I(x, y)$ (Assume it is smooth)

Consider a family of parametrized function $U(x, y, t)$ s.t

$$U(x, y, 0) = I(x, y).$$

Then $F(U) = g(U, t)$ a function of t .

$$\partial_t F(U) = \partial_t \int_{\Omega} |\nabla U|^2 dx dy = 2 \int_{\Omega} \langle \nabla U, \nabla U_t \rangle dx dy$$

Green's formula - $\int_{\Omega} \langle \Delta U, U_t \rangle dx dy + \int_{\partial\Omega} \frac{\partial U}{\partial \nu} \cdot U_t ds$

(Green's formula: $\int_{\Omega} u \cdot \Delta v dx dy = \int_{\partial\Omega} u \cdot \frac{\partial v}{\partial \nu} ds - \int_{\Omega} \nabla u \cdot \nabla v dx dy$)

So, if we choose

$$\begin{cases} U_t = \Delta U \\ \frac{\partial U}{\partial \nu} = 0 \end{cases} \quad \text{then } \partial_t F(U) \downarrow.$$

Experiment with Matlab.

Example 2: curve flow and level set method.

Background: Another way to average or smooth out.

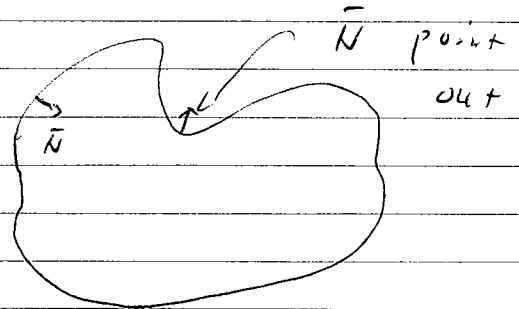
"Deforming a simple curve into a circle"
or an ellipse.

Curve flow problem (1985).

Plane curve $\bar{F}(p) := (x(p), y(p)) : p \in [0, 1] \rightarrow \mathbb{R}^2$

$\bar{N}(p)$: Unit inner normal vector

(roughly = second derivative of \bar{F})



Consider

$$(2) \quad \begin{cases} \bar{F}_t(p, t) = \tau \bar{N}(p, t) \\ \bar{F}(p, 0) = \bar{F}_0 \end{cases}$$

Rounder & Rounder.

① $\tau = k(p, t)$ ← curve shortening flow

② $\tau = k^{1/3}(p, t)$ ← Affine flow.

Level set method: curve flow for denoising.

Consider a gray level image

$$\phi(x, y) : \Omega \rightarrow \mathbb{R}^+$$

For given constant $c > 0$, then an contour curve (s) .

$$\phi(x, y) = c.$$

Now, we will move all contour curves simultaneously:

$$\phi[x(t), y(t)] = c$$

Then $\phi_x \cdot x_t + \phi_y \cdot y_t + \phi_t = 0.$

i.e. we want $\phi_t = - (x_t, y_t) (\phi_x, \phi_y)$

If all contour curves are moved by curve flow (2):

$$(x_t, y_t) = \tau \vec{N}^\perp = -\tau \left(\frac{\phi_x}{|\nabla\phi|}, \frac{\phi_y}{|\nabla\phi|} \right)$$

Then $\phi_t = \tau \cdot |\nabla\phi|!$

① $\tau = k$, $\phi_t = \frac{\phi_x^2 \phi_{yy} + \phi_y^2 \phi_{xx} - 2\phi_x \phi_y \phi_{xy}}{|\nabla\phi|^2}$

② $\tau = k^{1/3}$, $\phi_t = \left(\phi_x^2 \phi_{yy} + \phi_y^2 \phi_{xx} - 2\phi_x \phi_y \phi_{xy} \right)^{1/3}$

Since $k = \frac{\phi_x^2 \phi_{yy} + \phi_y^2 \phi_{xx} - 2\phi_x \phi_y \phi_{xy}}{|\nabla\phi|^2}$,

Experiments

```
function y=gaussian(F, items, step)

%Gaussian flow to smooth or blur image. MJ 7-05-2009.

[m,n]=size(F);
a=zeros(m,n);
b=zeros(m,n);
c=zeros(m,n);
d=zeros(m,n);

for k=1:items;

    f=F;
    a(1:m-1,:)=f(2:m,:);    %a(i,j)=f(i+1,j)
    a(m,:)=f(m,:);
    b(2:m,:)=f(1:m-1,:);%b(i,j)=f(i-1,j)
    b(1,:)=f(1,:);
    c(:,1:(n-1))=f(:,2:n);%c(i,j)=f(i,j+1)
    c(:,n)=f(:,n);
    d(:,2:n)=f(:,1:n-1);%d(i,j)=f(i,j-1)
    d(:,1)=f(:,1);

    fxx=a+b-2.*f; %fxx(i,j)=f(i+1,j)+f(i-1,j)-2f(i,j)
    fyy=c+d-2.*f; %fyy(i,j)=f(i,j+1)+f(i,j-1)-2f(i,j)

    ft=fxx+fyy;

    F=F+step.*ft; %stability: step<1/4.

end;

y=F;
```

```

function y=affine(f)
[m,n]=size(f);
a=zeros(m,n);
b=zeros(m,n);
c=zeros(m,n);
d=zeros(m,n);

for k=1:10;

    a(1:m-1,:)=f(2:m,:);    %a(i,j)=f(i+1,j)
    a(m,:)=f(m,:);
    b(2:m,:)=f(1:m-1,:);%b(i,j)=f(i-1,j)
    b(1,:)=f(1,:);
    c(:,1:(n-1))=f(:,2:n);%c(i,j)=f(i,j+1)
    c(:,n)=f(:,n);
    d(:,2:n)=f(:,1:n-1);%d(i,j)=f(i,j-1)
    d(:,1)=f(:,1);

    fx=a-f; %fx(i,j)=f(i+1,j)-f(i,j)
    fy=c-f; %fy(i,j)=f(i,j+1)-f(i,j)
    fxx=a+b-2.*f; %fxx(i,j)=f(i+1,j)+f(i-1,j)-2f(i,j)
    fyy=c+d-2.*f; %fyy(i,j)=f(i,j+1)+f(i,j-1)-2f(i,j)

    e(:,1:(n-1))=fx(:,2:n);%e(i,j)=fx(i,j+1)
    e(:,n)=fx(:,n);
    fxy=e-fx;%fxy(i,j)=(fx)y(i,j)=fx(i,j+1)-fx(i,j)

    %Euclidean shortening flow power=1
    %ft=(fx.^2.*fyy+fy.^2.*fxx-2.*fx.*fy.*fxy)./(fx.^2+fy.^2+0.00001);

    %affine flow power=1/3
    ft=nthroot((fy.^2.*fxx+fx.^2.*fyy-2.*fx.*fy.*fxy),3);

    f=f+0.1*ft;

end;
y=f;

```

For the first mathematical geodesic
 similar