

A Multi-Level Approach

To

**Context-preserving
Smooth Function Extension**

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Contents

1. Continuous function extension and applications
- 2. Motivation: Anisotropic diffusion (26 – 33)**
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Heat diffusion PDE models

Isotropic diffusion
(constant conductivity)

Anisotropic diffusions
(non-constant conductivity)

$$\left\{ \begin{array}{l} \frac{\partial}{\partial t} u(x,t) = \text{Div} (c \nabla u(x,t)) \\ u(x,0) = u_0(x) \end{array} \right. \quad \left\{ \begin{array}{l} \frac{\partial}{\partial t} u(\bullet,t) = \text{Div} (c(f(u),\bullet,t) \nabla u(\bullet,t)) \\ u(x,0) = u_0(x) \end{array} \right.$$

Perona-Malik anisotropic diffusion PDE

c as a function of only one variable

Perona and Malik 1990

$$u_t = \text{Div}(c(|\nabla u|)\nabla u)$$

$$\begin{cases} u_t = \text{Div}(c(|\nabla u|)\nabla u) & \text{in } D \\ \frac{\partial u}{\partial \mathbf{n}} = 0 & \text{on } \partial D \\ u(x, 0) = u_0(x) & \text{in } D \end{cases}$$

where D is the image domain

Refomulation of anisotropic diffusion (in terms of local coordinates)

$$\eta = \nabla u / |\nabla u|, \quad \phi(p) = p c(p), \quad p = |\nabla u|.$$

Diffusions along normal and tangential directions

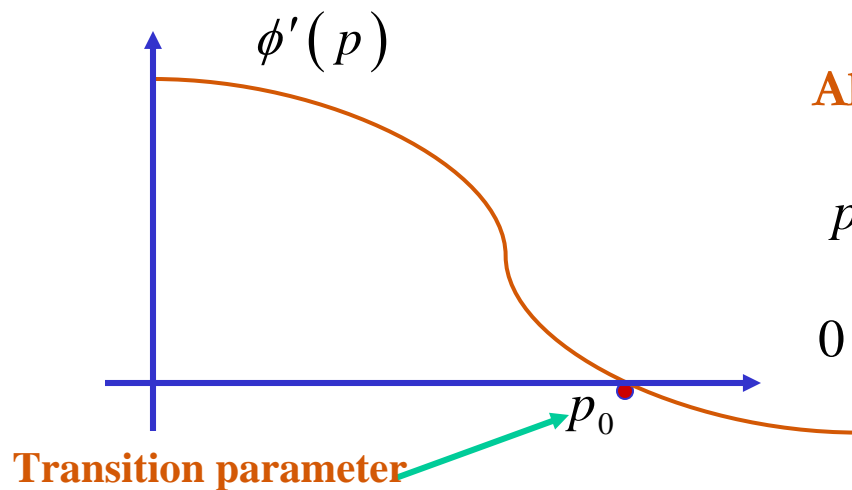
$$\left\{ \begin{array}{l}
 u_t = \boxed{\phi'(p) u_{\eta\eta}} + \boxed{c(p) u_{\tau\tau}} \\
 \frac{\partial u}{\partial n} = 0 \quad \text{on } \partial D \\
 u(x, 0) = u_0(x) \quad \text{in } D
 \end{array} \right.$$

$\phi'(p) = c(p) + p c'(p)$

Perona-Malik condition

$c(p) > 0$ decreases to 0 on $[0, \infty)$

$$\phi'(p) = c(p) + pc'(p) \rightarrow \begin{cases} \phi'(p) > 0, & \text{for } 0 \leq p < p_0 \\ \phi'(p) < 0, & \text{for } p > p_0 \end{cases}$$



Along the normal direction :

$p > p_0$ **Backward diffusion**

$0 \leq p < p_0$ **Forward diffusion**

Conductivity functions

$$c(p) = e^{-\frac{p^2}{2p_0^2}} \quad \text{Gaussian}$$

$$c(p) = \frac{p_0^2}{p_0^2 + p^2} \quad \text{Lorentz}$$

$$c(p) = \min\left(1, \frac{1}{p}\right) \quad \text{Normalized TV}$$

$$c(p) = 1 \quad \text{Isotropic}$$

p_0 transition parameter
 $p > p_0$ for backward diffusion

Anisotropic diffusion preserves image edges

Target (unknown)



Input image



Isotropic



Anisotropic



Lagged diffusion Model

For $j = 0, 1, \dots$, let $u^{j+1} = u^{j+1}(x, t)$

be the solution of

$$\begin{cases} u_t^{j+1} = \text{Div}(c(|\nabla u^j|)\nabla u^{j+1}) & \text{in } D \\ \frac{\partial u^{j+1}}{\partial \mathbf{n}} = 0 & \text{on } \partial D \\ u^{j+1}(x, 0) = u_0(x) & \text{in } D \end{cases}$$

with input $u^0(x, t) = u_0(x)$

Lagged diffusion model assures good approximation

Assuming “smooth” input u_0 , the convergence of $u^j = u^j(x, t)$ to the solution $u = u(x, t)$ of

$$\begin{cases} u_t = \text{Div}(c(|\nabla u|)\nabla u) & \text{in } D \\ \frac{\partial u}{\partial \mathbf{n}} = 0 & \text{on } \partial D \\ u(x, 0) = u_0(x) & \text{in } D \end{cases}$$

is uniform on $D \times [0, T]$ for any $T > 0$, with
(small “oh” of) **geometric** convergence rate.

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Problem statement

Simply connected domain Ω with holes D_1, \dots, D_m in \mathbb{R}^d , $d > 1$.
Consider the “Swiss cheese” domain

$$\tilde{\Omega} := \Omega \setminus (D_1 \cup \dots \cup D_m)$$

Given a sufficiently smooth function on Ω , but with missing portion on $D_1 \cup \dots \cup D_m$. The problem is to recover the missing portion.

Since our approach is local, it is sufficient to consider a single hole $D \subset \Omega$ so that $\tilde{\Omega} := \Omega \setminus D$

Diffusion and data propagation

Multi-level heat kernels:

$$G_j(x, y)$$

Diffusion operators:

$$(T_j f)(x) = \int_D f(y) G_j(x, y) dy$$

Data propagation operators:

$$P_j(d_j F)(x) = - \int_{\partial D} (d_j F)(y) g_j(x, y) ds$$

Detail-extension operators:

$$E_j = T_0 \cdots T_{j-1} P_j, \quad j = 1, \dots, n$$

Wavelet details:

$$w_j = E_j(d_j(F)), \quad j = 1, \dots, n$$

Lagged anisotropic conductivity

Let

$$c_j(x) := c(|\nabla u^j(x)|) \quad \text{or} \quad c(|\nabla s * u^j(x)|), \quad x \in D \subset \mathbb{R}^d$$

for some lowpass function s , such as the Gaussian,

with

$$c_{-1}(x) = 1$$

and consider the differential operators

$$(L_j f)(x) = \nabla \cdot (c_{j-1}(x) \nabla f(x)), \quad x \in D$$

Diffusion, propagation, and detail-extension

Green's function:
$$\begin{cases} (L_j)G_j(x, y) = \delta(x, y), & x, y \in D \\ G_j(x, y)|_{y \in \partial D} = 0, & x \in D \end{cases}$$

“Diffusion” operator:
$$(T_j f)(x) = \int_D f(y)G_j(x, y)dy \quad (1)$$

Propagation kernel:
$$g_j(x, y) := -c_{j-1}(y) \left(\frac{\partial}{\partial \mathbf{N}} G_j(x, \cdot) \right) (y) \quad (2)$$

Propagation operator:
$$(P_j v)(x) = \int_{\partial D} v(y(s))g_j(x, y(s))ds \quad (3)$$

Detail-extension operator:
$$E_j = T_0 \cdots T_{j-1}P_j, \quad j = 1, \dots, n \quad (4)$$

The ML approach

1. Apply the propagation operator P_0 defined in (3) with $j = 0$ to transport the boundary data $d_0(F) = F|_{\partial D}$ to D to construct the lowest resolution level (called thumb-nail).
2. Compute mixed derivatives $d_1(F), \dots, d_n(F)$ of increasing order, using data $F(x)$ from the exterior of D .
3. Extend the mixed derivatives data $d_1(F), \dots, d_n(F)$ by applying the operators E_1, \dots, E_n defined in (4) to D .

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Extension of Green's formula

Lemma 1. *For sufficiently smooth functions $f, h, c_{j-1} \in C(D)$ with $h|_{\partial D} = 0$,*

$$\int_D f(y)(L_j h)(y)dy = \int_D h(y)(L_j f)(y)dy - \int_{\partial D} c_{j-1}(y)f(y)\frac{\partial}{\partial \mathbf{N}} h(y)ds,$$

where y is parametrized in terms of s .

Boundary data, wavelets, and MRA solution

Choice of data: $d_j(F) := L_{j-1} \cdots L_0 F$ on ∂D

Lowest resolution level: $u_0 = P_0(F|_{\partial D})$

Wavelet details: $w_j = E_j(d_j(F)), \quad j = 1, \dots, n$

MRA solution: $u_n := u_0 + w_1 + \cdots + w_n$

Continuity preservation: $u_n|_{\partial D} = u_0|_{\partial D} = F|_{\partial D}, \quad n = 1, 2, \dots$

Error formula

Theorem 1 (ACHA, 08) *Let F be a sufficiently smooth function in Ω*

but with missing portion $F_D := F|_D$, $D \subset \Omega$

Then the error of ML recovery of the lost data is given by

$$F_D(x) - u_n(x) = (T_0 \cdots T_n)(L_n \cdots L_0 F_D)(x), \quad x \in D.$$

Error bound

In particular,

$$\|F_D - u_n\|_D \leq C_n \prod_{j=0}^n \|T_j\|,$$

where

$$C_n := \|(L_n \cdots L_0)F_D\|_D$$

and

$$\|T_j\| := \sup_{x \in D} \int_D |G_j(x, y)| dy.$$

Proof: Lowest resolution

As a consequence of Lemma 1,

$$\begin{aligned} F_D(x) &= \int_D F_D(y) L_0 G_0(x, y) dy \\ &= \int_D (L_0 F_0)(y) G_0(x, y) dy - \int_{\partial D} F(y) c_{-1}(y) \frac{\partial}{\partial \mathbf{N}} G_0(x, y) ds \\ &= (T_0(L_0 F_D))(x) + u_0(x) \end{aligned}$$

or

$$F_D(x) - u_0(x) = (T_0(L_0 F_D))(x)$$

Proof: Higher resolutions

For each $j = 1, \dots, n$, an application of Lemma 1 also yields:

$$\begin{aligned} f_j(x) &:= (L_{j-1} \cdots L_0 F_D)(x) = \int_D f_j(y) L_j G_j(x, y) dy \\ &= \int_D (L_j f_j)(y) G_j(x, y) dy - \int_{\partial D} f_j(y) g_j(x, y) ds \\ &= (T_j(L_j \cdots L_0 F_D))(x) - \int_{\partial D} (d_j F)(y) g_j(x, y) ds \\ &= (T_j(L_j \cdots L_0 F_D))(x) + P_j(d_j F)(x). \end{aligned}$$

Proof: Wavelet details

Hence, by applying the operator $T_0 \cdots T_{j-1}$ to each term and the definition (4), we have

$$(T_0 \cdots T_{j-1})(L_{j-1} \cdots L_0 F_D)(x) = (T_0 \cdots T_j)(L_j \cdots L_0 F_D)(x) + (E_j(d_j F))(x),$$

which, by telescoping, yields

$$(T_0(L_0 F_D))(x) - \sum_{j=1}^n (E_j(d_j F))(x) = (T_0 \cdots T_n)(L_n \cdots L_0 F_D)(x).$$

Recall that

$$F_D(x) - u_0(x) = (T_0(L_0 F_D))(x),$$

$$w_j = E_j(d_j(F)), \quad j = 1, \dots, n$$

Extension to Manifolds

In a joint paper with H. Mhaskar (ACHA, 09)

Smooth function extension on manifolds with
context recovery

(not discussed in this talk)

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Special case: Isotropic setting

Conductivity: $c(p) = 1$

Differential operators: $L_j = \Delta$

Boundary data: $d_0(F) = F|_{\partial D}$ and $d_j(F) = \Delta^j F|_{\partial D}$, $j = 1, \dots, n$

Wavelets: $w_j = E_j(d_j(F))$, $j = 1, \dots, n$

MRA solution: $u_n := u_0 + w_1 + \dots + w_n$

Isotropic setting: Peano kernels

Lemma 2. The “Peano” kernels

$$K_1(x, y) := G^\Delta(x, y)$$

$$K_{j+1}(x, y) := \int_{D^j} G^\Delta(x, y_j) \cdots G^\Delta(x_k, y_{k-1}) \cdots G^\Delta(y_1, y) dy_1 \cdots dy_j$$

$j = 1, 2, \dots, n$, satisfy:

$$\Delta_x K_{j+1}(x, y) = K_j(x, y)$$

$$\Delta_x^j K_j(x, y) = \delta(x, y)$$

Isotropic setting: Formulas

Propagation operator: $(Pf)(x) := \int_{\partial D} f(y) \frac{\partial}{\partial \mathbf{N}} G^\Delta(x, y) ds$

Ground level: $u_0(x) = P(F|_{\partial D})(x)$

Wavelets: $w_j(x) = \int_D K_j(x, y) P(\Delta^j F|_{\partial D})(y) dy$

MRA: $u_n(x) = u_0(x) + w_1(x) + \dots + w_n(x)$

Error (Remainder): $F(x) - u_n(x) = \int_D K_{n+1}(x, y) \Delta^{n+1} F(y) dy$

Isotropic setting: Interpolation properties

Theorem 2. For $k = 0, \dots, n$ and $j = 1, \dots, n$, the wavelet components satisfy the boundary conditions:

$$\Delta^k w_j \Big|_{\partial D} = \delta_{j,k} \Delta^j F \Big|_{\partial D}$$

Hence, the continuous function extension

$$u_n(x) = u_0(x) + w_1(x) + \dots + w_n(x)$$

satisfies the interpolating property:

$$u_n \Big|_{\partial D} = F \Big|_{\partial D} \quad \Delta^j u_n \Big|_{\partial D} = \Delta^j F \Big|_{\partial D}, \quad j = 1, \dots, n$$

Smooth harmonic inpainting

T. Chan and J. Shen

Dirichlet problem:
$$\begin{cases} \Delta u^h = 0 & \text{in } D, \\ u^h|_{\partial D} = F|_{\partial D} & \text{on } \partial D \end{cases}$$

Linear inpainting:
$$\|F_D - u^h\|_D = O(\epsilon^2)$$

Cubic inpainting:
$$\|F_D - u^{bh}\|_D = O(\epsilon^4)$$

Error parameter:
$$\text{diam } (D) = 2\epsilon > 0$$

Smooth bi-harmonic inpainting

T. Chan and J. Shen

Bi-harmonic

inpainting function:

$$u^{bh} := u^h + u^a$$

Detail component:

$$\begin{cases} \Delta u^a = \tilde{u}^h & \text{in } D, \\ u^a|_{\partial D} = 0 \end{cases}$$

where

$$\begin{cases} \Delta \tilde{u}^h = 0 & \text{in } D, \\ \tilde{u}^h|_{\partial D} = \Delta F|_{\partial D} & \text{on } \partial D \end{cases}$$

MRA extension and sharp error estimates

Linear inpainting: $u^h = u_0$

Cubic inpainting: $u^{bh} = u^h + u^a = u_0 + w_1$

Our continuous function extension is valid for
any number of levels $u_n := u_0 + w_1 + \cdots + w_n$

From Theorem 1, we also have sharp error
estimates:

$$\|F_D - u_n\|_D \leq \|\Delta^{n+1} F_D\|_D \|T^\Delta\|^{n+1}$$

Improvement and extension of Chan-Shen error estimates

Error parameter: $\text{diam } (D) = 2\epsilon > 0$

Green's function
of Laplacian: $\|T^\Delta\| := \sup_{x \in D} \int_D |G^\Delta(x, y) dy| = \sup_{x \in D} \left| \int_D G^\Delta(x, y) dy \right|$

For $D = D_\epsilon := \{|x - x_0| < \epsilon\} \subset \mathbb{R}^d$

$$\int_D G^\Delta(x, y) dy = \frac{|x - x_0|^2 - \epsilon^2}{2d}$$

Hence,

$$\|F_D - u_n\|_D \leq \|\Delta^{n+1} F_D\|_D \left(\frac{\epsilon^2}{2d}\right)^{n+1} \quad n = 0, 1, \dots$$

This holds for $D \subset D_\epsilon$:

Two-dimensional setting

Green's function: $G^\Delta(x, y) = \frac{1}{2\pi} \ln |\phi_x(y)|$

where $\phi_x(y), x \in D \subset \mathbb{R}^2$ is the conformal map of D onto the unit disk with center at $\mathbf{0}$, such that $\phi_x(x) = 0$ and $\phi'_x(x) > 0$ so that

$$\int_D G^\Delta(x, y) dy$$

can be estimated more directly.

Thank You

Questions and comments?