

Diffusion Kernels on Graphs

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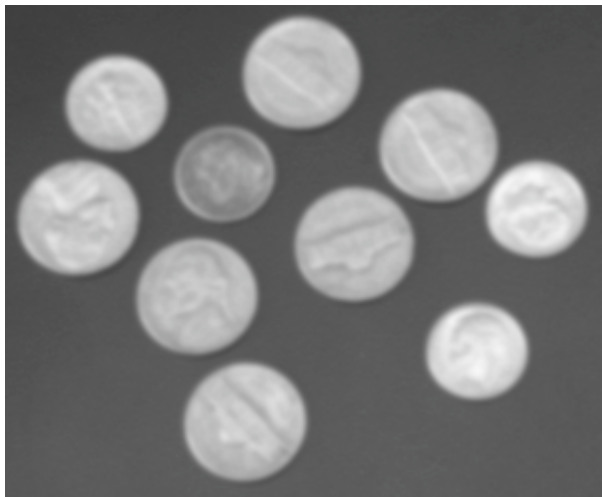
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February 9, 2007

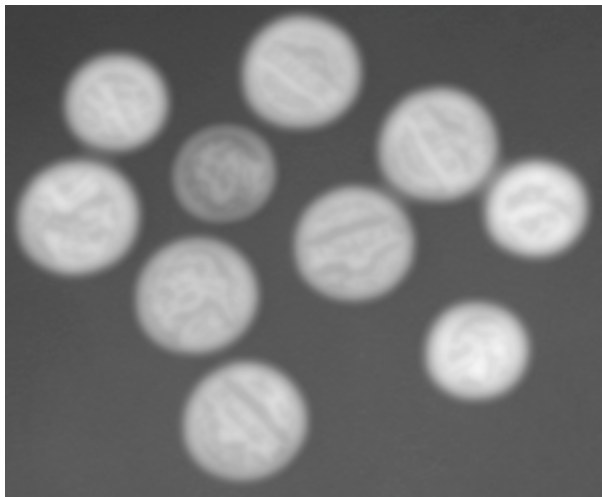
Outline

- ▶ Examples of Diffusion
- ▶ Motivations for a heat equation when the space variable Belongs at the vertices of a graph
- ▶ heat equation when the space variable belongs \mathbf{R}^2
- ▶ Heat equation over a graph
- ▶ Application: Models for natural language

Example of diffusion



Example of diffusion



Motivation for a Heat equation over a graph. Image processing

Image smoothing

Image: function $g : S \mapsto \mathbf{R}$, S is a discrete set of pixels.

Noisy image: $g(s) + \epsilon(s)$

Recover g by diffusing.

Idea: Build a graph. The vertices are S . Define the edges ... Start at $g(s) + \epsilon(s)$ and diffuse and stop ...

Motivation for a Heat equation over a graph. Language processing

Closed vocabulary V , $\#V = K \approx 10^5$

Training set of words x_1, \dots, x_m , counts $n(w_1), \dots, n(w_K)$

Want to build a probability mass function π over the words of V

i.e $\pi(w_k) \geq 0$ and $\sum_{k=1}^K \pi(w_k) = 1$

What is the probability assigned to unseen words ? ...

Idea: Build a graph. Vertices = V , Define the edges ... start at

$\pi_0(w_k) = m^{-1}n(w_k)$ and diffuse and stop ...

Heat equation in \mathbf{R}^2

$x = (x_1, x_2) \in \mathbf{R}^2, y = (y_1, y_2) \in \mathbf{R}^2, t \geq 0, \alpha > 0$

$K_t(x, y)$ is the temperature at time t at x when starting at time $t = 0$ with all the heat concentrated at y . It is called a diffusion kernel.

$$\text{for all } x, \text{ for all } t > 0, \frac{\partial}{\partial t} K_t(x, y) = \alpha \Delta K_t(x, y)$$

Δ stands for Laplacian.

$$\Delta K_t(x, y) = \frac{\partial^2}{\partial^2 x_1} K_t((x_1, x_2), y) + \frac{\partial^2}{\partial^2 x_2} K_t((x_1, x_2), y)$$

Heat equation in \mathbb{R}^2

Without restricting the domain, the solution is given by

$$K_t(x, y) = \frac{1}{4\pi\alpha t} \exp\left(-\frac{1}{4\alpha t} ((x_1 - y_1)^2 + (x_2 - y_2)^2)\right)$$

$K_t(x, y)$ is the density of a

$$N(y, 2\alpha t Id)$$

If now the temperature at time 0 is given by $g(x)$ then the solution of the heat equation is the convolution

$$\int \int K_t(x, y) g(y) dy$$

Discretization of the Laplacian

$x = (x_1, x_2) \in \mathbf{R}^2$, $f : \mathbf{R}^2 \mapsto \mathbf{R}$

$$\begin{aligned}\frac{\partial}{\partial x_1} f(x_1, x_2) &\approx \frac{1}{h} \left(f(x_1 + \frac{h}{2}, x_2) - f(x_1 - \frac{h}{2}, x_2) \right) \\ \frac{\partial^2}{\partial^2 x_1} f(x_1, x_2) &\approx \frac{1}{h} \left(\frac{\partial}{\partial x_1} f(x_1 + \frac{h}{2}, x_2) - \frac{\partial}{\partial x_1} f(x_1 - \frac{h}{2}, x_2) \right) \\ &\approx \frac{1}{h} \left(\frac{1}{h} (f(x_1 + h, x_2) - f(x_1, x_2)) - \right. \\ &\quad \left. \frac{1}{h} (f(x_1, x_2) - f(x_1 - h, x_2)) \right) \\ &\approx \frac{1}{h^2} (f(x_1 + h, x_2) + f(x_1 - h, x_2) - 2f((x_1, x_2))) \\ \frac{\partial^2}{\partial^2 x_2} f(x_1, x_2) &\approx \frac{1}{h^2} (f(x_1, x_2 + h) + f(x_1, x_2 - h) - 2f((x_1, x_2)))\end{aligned}$$

Discretization of the Laplacian

$$\begin{aligned}\Delta f(x_1, x_2) &= \frac{\partial^2}{\partial^2 x_1} f(x_1, x_2) + \frac{\partial^2}{\partial^2 x_2} f(x_1, x_2) \\ &= \frac{1}{h^2} (f(x_1 + h, x_2) + f(x_1 - h, x_2) - 2f((x_1, x_2))) + \\ &\quad \frac{1}{h^2} (f(x_1, x_2 + h) + f(x_1, x_2 - h) - 2f((x_1, x_2)))\end{aligned}$$

Define $\mathcal{V}(x) = \{(x_1 + h, x_2), (x_1 - h, x_2), (x_1, x_2 - h), (x_1, x_2 + h)\}$ and $d(x) = \#\mathcal{V}(x)$ then

$$\Delta f(x) = \frac{1}{h^2} \left(\left(\sum_{y \in \mathcal{V}(x)} f(y) \right) - d(x)f(x) \right)$$

Heat equation over a graph

$G(V, E)$ a non oriented graph.

$V = \{x_1, \dots, x_n\}$ is the finite set of vertices.

$E \subset V \times V$ is the set of edges. If $(x, y) \in E$, we denote $x \sim y$. Assume no edge from a vertex to itself. Assume G is connected.

The degree of $x \in V$ is $d(x) = \sum_{y \in V} \delta(x \sim y)$

$f : V \mapsto \mathbf{R}$ can be seen as a function or as a vector $(f(x_1), \dots, f(x_n))^T$

$H : V \times V \mapsto \mathbf{R}$ can be seen as a function or as a $n \times n$ matrix.

Define the Laplacian (choose $h = 1$)

$$\begin{aligned}\Delta f(x) &= \left(\sum_{y \in \mathcal{V}(x)} f(y) \right) - d(x)f(x) \\ &= \left(\sum_{y: y \sim x} f(y) \right) - d(x)f(x) \\ &= \sum_{y \in \mathcal{V}} (f(y)\delta(y \sim x) - d(y)f(y)\delta(x = y)) \\ &= \sum_{y \in \mathcal{V}} (\delta(y \sim x) - d(y)\delta(x = y)) f(y) \\ &= \sum_{y \in \mathcal{V}} H(x, y)f(y) \\ &= Hf(x)\end{aligned}$$

Laplacian

$$H(x, y) = \delta(y \sim x) - d(y)\delta(x = y)$$

$$H = A - D$$

$A(x, y) = \delta(x \sim y)$ is the adjacency matrix of G

$D(x, y) = d(x)\delta(x = y)$ is the degree matrix. D is diagonal.

Heat Equation

$x, y \in V, t \geq 0$.

$K_t(x, y)$ is the temperature at x at time t when starting with a unit temperature at y at time 0.

$K_0(x, y) = \delta(x = y)$ which in matrix notation is $K_0 = Id$

We define the heat equation for a fixed $y \in V$ as:

$$\text{for each } x \in V, \text{ for each } t > 0, \frac{\partial}{\partial t} K_t(x, y) = HK_t(x, y)$$

Notate $u_t(x) = K_t(x, y)$

$$\begin{aligned} \frac{\partial}{\partial t} u_t(x) &= \sum_{z \in V} H(x, z) u_t(z) \\ &= \left(\sum_{z: z \sim x} u_t(z) \right) - d(x) u_t(x) \\ &= d(x) \left(\left(\frac{1}{d(x)} \sum_{z: z \sim x} u_t(z) \right) - u_t(x) \right) \end{aligned}$$

Claims:

The heat equation admits a unique solution $K_t = e^{tH}$

$$e^{tH} = Id + tH + \frac{t^2}{2!}H^2 + \frac{t^3}{3!}H^3 + \dots$$

$$e^{tH} = \lim_{k \rightarrow +\infty} \left(Id + \frac{t}{k}H \right)^k$$

Starting with a temperature $\pi(x)$, $x \in V$, the solution to the heat equation is $K_t\pi$

If for all x , $\pi(x) \geq 0$ and $\sum_{x \in V} \pi(x) = 1$ then for all $x \in V$ and all $t > 0$, $K_t\pi(x) > 0$ and $\sum_{x \in V} K_t\pi(x) = 1$

Markov Chain interpretation ...

Examples

- ▶ Complete graph with n vertices. $x \sim y \iff x \neq y$

$$\begin{aligned}K_t(x, y) &= \frac{1}{n} (1 + (n-1)e^{-nt}) \text{ if } x \neq y \\ &= \frac{1}{n} (1 - e^{-nt}) \text{ if } x = y\end{aligned}$$

- ▶ Vertices are binary strings of length n .

$$x \sim y \iff \text{Hamming}(x, y) = 1$$

$$K_t(x, y) = \frac{1}{2^n} (1 + e^{-2t})^n (\tanh(t))^{H(x, y)}$$

- ▶ small graphs. Diagonalize H

Application to language modeling

Closed vocabulary V , $\#V = K \approx 10^5$

Training set of words x_1, \dots, x_m , counts $n(w_1), \dots, n(w_K)$

Want to build a probability mass function π over the words of V

i.e $\pi(w_k) \geq 0$ and $\sum_{k=1}^K \pi(w_k) = 1$ From observed counts

$\pi_0(w_k) = m^{-1}n(w_k)$ Vertices = V ,

- ▶ Choose the complete graph over V . $x \sim y \iff x \neq y$. Start at π_0
Then ... for all $w \in V$, $K_t \pi(w) = (1 - \lambda(t))\pi_0(w) + \lambda(t)\frac{1}{K}$
- ▶ Choose $v \sim w \iff |n(v) - n(w)| \leq 1$, compute $(Id + \frac{t}{3}H)^3 \pi_0$ with $t = \frac{1}{K}$ yields fast and competitive results.

Thank you