Week 4

Chapter 4

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4.1 Transition densities and Laplacian

We can recall that $\mathbf{P}(X_1 = y) = \mathcal{P}(x, y) = \frac{\mu_{xy}}{\mu_x}$.

Definition 4.1.1. The transition density w.r.t weights μ of a random walk $\{X_n\}$ is given by:

$$p_n(x,y) = \frac{\mathbf{P}^x(X_n = y)}{\mu_y} := \frac{\mathcal{P}_n(x,y)}{\mu_y} \qquad \mu \ge 1$$

$$\begin{array}{l} p_0(x,y) = \frac{1_{\{x\}}(y)}{\mu_y} = \frac{1_{\{x\}}(y)}{\mu_x} \\ p(x,y) \equiv p_1(x,y) = \frac{\mu_{xy}}{\mu_x \mu_y} \end{array}.$$

Lemma 4.1.1. Let p_n be the transition densities of $\{X_n\}_{n\geq 0}$

1.
$$p_{n+m}(x,y) = \sum_{z \in V} p_n(x,z) p_m(z,y) \mu_z$$

2.
$$\forall x, y \in V, p_n(x, y) = p_n(y, x)$$

3.
$$\forall x, y \in V, \sum_{z \in V} p_n(x, z) \mu_z = 1 = \sum_{z \in V} p_n(z, y) \mu_z$$

Proof. 1.

$$p_{n+m}(x,y) = \frac{\mathbf{P}^{x}(X_{n+m} = y)}{\mu_{y}}$$

$$= \sum_{z \in V} \frac{\mathbf{P}^{x}(X_{n+m} = y, X_{n} = z)}{\mu_{y}}$$

$$= \frac{1}{\mu_{y}} \sum_{z \in V} \sum_{0 \le i < n+m, x_{i} \in V} 1_{\{x\}}(x_{0}) \prod_{i=0}^{n-1} \mathcal{P}(x_{i}, x_{i+1}) 1_{\{z\}}(x_{n}) \prod_{i=n}^{n+m} \mathcal{P}(x_{i}, x_{i+1}) 1_{\{y\}}(x_{n+m})$$

$$\stackrel{\text{H1}}{=} \frac{1}{\mu_{y}} \sum_{z \in V} \mathbf{P}^{x}(X_{n} = z) \mathbf{P}^{z}(X_{m} = y)$$

$$= \frac{1}{\mu_{y}} \sum_{z \in V} p_{n}(x, z) \mu_{z} p_{m}(z, y) \mu_{y}$$

$$= \sum_{z \in V} p_{n}(x, z) p_{m}(z, y) \mu_{z}$$

2.

$$p_n(x,y) = \frac{\mathbf{P}^x(X_n = y)}{\mu_y} = \frac{\mathbf{P}^y(X_n = x)}{\mu_x} = p_n(y,x)$$

The second equality is obtained by applying the Detailed Balance equations.

3.

$$\sum_{z \in V} p_n(x, z) \mu_z = \sum_{z \in V} \mathbf{P}^x (X_n = z) = 1$$
$$\sum_{z \in V} p_n(z, y) \mu_z = \sum_{z \in V} p_n(y, z) \mu_z = \sum_{z \in V} \mathbf{P}^y (X_n = z) = 1$$

4.2 Function Spaces

Definition 4.2.1.

$$C(V) = \{f : V \to \mathbb{R}\} = \mathbb{R}^V$$

$$Co(V) = \{f : V \to \mathbb{R}, f \neq 0 \text{ on finitely many points}\}$$

$$C_+(V) = \{f : f \in C(V), f \geq 0\}$$

$$Supp(f) = \{x : x \in V, f(x) \neq 0\}$$

Definition 4.2.2. We define the **norm** of a function as the following

$$\forall p \in [1, \infty), \|f\|_p = \left(\sum_{x \in V} |f(x)|^p \mu_x\right)^{\frac{1}{p}}$$
$$\|f\|_{\infty} = \sup\{|f(x)| : x \in V\}$$

f is said to be L^p on the graph with vertex set V and weights μ if and only if f is a function defined on the vertex set, V and its p-norm is finite everywhere.

$$f \in L^p(V, \mu) \iff f \in C(V) \text{ and } ||f||_p < \infty$$

Definition 4.2.3. We define an inner product on the $L^2(V, \mu)$ space in the following way

$$\langle f, g \rangle = \sum_{x \in V} f(x)g(x)\mu_x$$

$$\mathbf{E}[f(X_n)] = \sum_{x \in V} f(z) \mathbf{P}^x (X_n = z)$$
$$= \sum_{x \in V} f(z) p_n(x, z) \mu_z$$
$$= \langle f, p_n(x, .) \rangle$$

which brings us to define a new function

Definition 4.2.4. $\mathcal{P}_n: C(V) \to C(V)$ given by

$$\mathcal{P}_n f(x) = \sum_{x \in V} f(z) p_n(x, z) \mu_z = \langle f, p_n(x, .) \rangle$$

where $\Delta: C(V) \to C(V)$ as an "operation" on C(V) is

$$\Delta = P - I$$

We write $\mathcal{P}_1 f(x)$ as $\mathcal{P} f(x)$ and proceed to look at computations and lemmas involving $\mathcal{P} f$.

Lemma 4.2.1.

$$\forall x \in V, \ \mathcal{P}f(x) - f(x) = \Delta f(x)$$

Proof.

$$\mathcal{P}f(x) - f(x) = \sum_{x \in V} f(z)p(x, z)\mu_z - f(x)$$

$$\stackrel{*}{=} \sum_{x \in V} p(x, z)\mu_z(f(z) - f(x))$$

$$= \sum_{x \in V} \frac{\mu_{xz}}{\mu_x \mu_z} \mu_z(f(z) - f(x))$$

$$= \frac{1}{\mu_x} \sum_{x \in V} \mu_{xz}(f(z) - f(x))$$

$$= \Delta f(x)$$

Corollary 4.2.1.

$$\Delta f = 0 \iff f(x) = \mathcal{P}f(x) = \mathbf{E}^x[f(X_1)]$$

Definition 4.2.5. We define a function $A: C(V) \to C(V)$ as

$$||A||_{n\to n} = \sup\{||Af||_n : ||f||_n \le 1\}$$

Proposition 4.2.1. 1. $\mathcal{P}1 = 1$ where $1(x) = 1 \ \forall x \in V$

- 2. $|\mathcal{P}f| \leq \mathcal{P}|f|$ where $f \in C(V)$
- 3. $\|\mathcal{P}\|_{p\to p} \le 1$ $\|\Delta\|_{p\to p} \le 2$ where $p \in [1, \infty) \cup \{\infty\}$

^{*} is left as an exercise and can be proved using property 2 from (4.1.1)

$$\mathcal{P}1(x) = \sum_{x \in V} p(x, z)\mu_z = 1 = 1(x)$$

2.

$$|\mathcal{P}f(x)| = \left| \sum_{x \in V} f(z)p(x, z)\mu_z \right|$$

$$\leq \sum_{x \in V} |f(z)| p(x, z)\mu_z$$

$$= \mathcal{P}|f|(x)$$

3.

$$\|\mathcal{P}f\|_{p}^{p} = \sum_{x \in V} |\mathcal{P}f(x)|^{p} \mu_{x}$$

$$= \sum_{x \in V} \left| \sum_{z \in V} f(z)p(x,z)\mu_{z} \right|^{p} \mu_{x}$$

$$\stackrel{*}{\leq} \sum_{x \in V} \left(\sum_{z \in V} |f(z)|^{p} p(x,z)\mu_{z} \right) \left(\sum_{z \in V} 1^{q} p(x,z)\mu_{z} \right) \mu_{x}$$

$$= \sum_{x \in V} \left(\sum_{z \in V} |f(z)|^{p} p(x,z)\mu_{z} \right) \mu_{x}$$

$$\stackrel{**}{=} \sum_{z \in V} |f(z)|^{p} \mu_{z}$$

$$= \|f\|_{p}$$

$$\implies \|\mathcal{P}\|_{n \to n} \leq 1$$

$$(4.1)$$

where $\frac{1}{p} + \frac{1}{q} = 1$

We leave the proofs of the following as exercises

*, which can be proved using Holder's inequality, ** and the $p = \infty$ case

$$\begin{split} \|\Delta f\|_p^p &= \|\mathcal{P}f - f\|_p^p \\ &\leq (\|\mathcal{P}f\|_p + \|f)\|_p)^p \\ &\leq 2^{p-1} (\|\mathcal{P}f\|_p^p + \|f\|_p^p) \\ &\leq (2\|f\|_p)^p \qquad \qquad [\because \|\mathcal{P}f\|_p \leq \|f\|_p] \\ \Longrightarrow \|\Delta\|_{p \to p} \leq 1 \end{split}$$

The final inequality is obtained from (4.1).

Proposition 4.2.2. \mathcal{P} is self-adjoint on $L^2(V,\mu)$

$$\forall f, g \in L^2(V, \mu), \langle \mathcal{P}f, g \rangle = \langle f, \mathcal{P}g \rangle$$

Proof.

$$\begin{split} \langle \mathcal{P}f,\;g\rangle &= \sum_{x\in V} \mathcal{P}f(x)g(x)\mu_x\\ &= \sum_{x\in V} (\sum_{z\in V} f(z)p(x,z)\mu_z)g(x)\mu_x\\ &\stackrel{Ex}{=} \sum_{z\in V} f(z)\mu_z \sum_{x\in V} p(z,x)g(x)\mu_x\\ &= \sum_{z\in V} f(z)\mathcal{P}g(z)\mu_z\\ &= \langle f,\;\mathcal{P}g\rangle \end{split}$$

4.3 Dirichlet forms

Definition 4.3.1. We define the quadratic form on $L^2(V,\mu)$, \mathcal{E} as

$$\mathcal{E}(f,g) = \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (f(x) - f(y))(g(x) - g(y))\mu_{xy}$$

whenever the series converges absolutely.

Theorem 4.3.1 (Discrete Green's Theorem). $\forall f, g \in C(V)$,

$$\sum_{x \in V} \sum_{y \in V} |f(x) - f(y)| |g(x)| \mu_x < \infty$$

$$\implies \mathcal{E}(f, g) = -\langle \Delta f, g \rangle$$

We present an application of (4.3.1)

Lemma 4.3.1. Let (Γ, μ) be a weighted graph such that $\mu(V) < \infty$. Then, (Γ, μ) is **recurrent**.

Proof. Fix $Z \in V$ Define $\Phi: V \to \mathbb{R}$ where $\Phi(x) := \mathbf{P}^x(\mathcal{T}_z = \infty)$

- 1. Firstly observe that $\Phi(z) = \mathbf{P}^z(\mathcal{T}_z = \infty) = 0$
- 2. $\forall n \geq 1, x \neq z$ $\Phi_n(x) := \mathbf{P}^x(\mathcal{T}_z = n) = \sum_{u \in V} \mathcal{P}(x, u) \Phi_{n-1}(x)$

This holds true from a simple logical argument. Starting from x, hitting z in n steps is equivalent to jumping from x to some vertex u and hitting z in n-1 steps.

3.
$$1 - \Phi(x) = \mathbf{P}^x(\mathcal{T}_z < \infty) = \sum_{n=0}^{\infty} \mathbf{P}^x(\mathcal{T}_z = n) = \sum_{n=1}^{\infty} \Phi_n(x)$$

4. $\Phi \equiv 0$

$$\sum_{n=1}^{k} \Phi_n(x) = \sum_{n=1}^{k} \sum_{u \in V} \mathcal{P}(x, u) \Phi_{n-1}(u)$$

$$\implies \sum_{n=1}^{k} \Phi_n(x) = \sum_{u \in V} \mathcal{P}(x, u) \sum_{n=1}^{k} \Phi_{n-1}(u)$$

$$\implies \sum_{n=1}^{\infty} \Phi_n(x) = \sum_{u \in V} \mathcal{P}(x, u) \sum_{n=1}^{\infty} \Phi_{n-1}(u)$$

$$\implies 1 - \Phi(x) = \sum_{u \in V} \mathcal{P}(x, u) (1 - \Phi(u))$$

$$\implies 1 - \Phi(x) = \sum_{u \in V} p(x, u) (1 - \Phi(u)) \mu_u$$

$$\implies 1 - \Phi = \mathcal{P}(1 - \Phi)$$

$$\implies \Delta(1 - \Phi) = 0$$

Then, by theorem (4.3.1),

$$\mathcal{E}(1 - \Phi, 1 - \Phi) = \langle \Delta(1 - \Phi), 1 - \Phi \rangle = 0$$

$$\implies \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (\Phi(x) - \Phi(y))^2 \mu_{xy} = 0$$

$$\implies \Phi(x) = \Phi(y) \qquad \forall \ x, y \in V$$

$$\implies \Phi \text{ is constant}$$

$$\implies \Phi \equiv 0$$

The last equality holds as $\Phi(z) = 0$.

Since, $\Phi \equiv 0$ for arbitrary z, (Γ, μ) is recurrent.

To proof theorem (4.3.1), we start with some prerequisites.

Definition 4.3.2.

$$\mathcal{H}^{2}(V) = \{ f : f \in C(v), \ \mathcal{E}(f, f) < \infty \}$$
$$\|f\|_{\mathcal{H}^{2}} = \sqrt{\mathcal{E}(f, f) + f^{2}(\rho)} \quad \text{for some fixed } \rho \in V$$

Proposition 4.3.1. Let (Γ, μ) be a graph satisfying properties, H1 and H2.

1.
$$|f(x) - f(y)| \le \frac{1}{\sqrt{\mu_{xy}}} \sqrt{\mathcal{E}(f, f)}$$
 $\forall x \sim y$

2. $\mathcal{E}(f,f) = 0 \iff f \text{ is constant}$

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3.
$$f \in L^2 \implies \mathcal{E}(f, f) \le 2 \|f\|_2^2$$

Proof. 1. If $\mathcal{E}(f, f) = \infty$, then we are done Let $\mathcal{E}(f, f)$ be finite

$$\mathcal{E}(f, f) = \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (f(x) - f(y))^2 \mu_{xy}$$
$$\geq (f(x) - f(y))^2 \mu_{xy} \quad \forall x, y \in V$$

2. The forward direction is left as an exercise. The reverse direction follows from the definition.

3.

$$\mathcal{E}(f,f) \leq \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (f(x) - f(y))^2 \mu_{xy}$$

$$\leq \sum_{x \in V} \sum_{y \in V} (|f(x)|^2 + |f(y)|^2) \mu_{xy}$$

$$\stackrel{Ex}{=} \sum_{x \in V} |f(x)|^2 \mu_x + \sum_{y \in V} |f(y)|^2 \mu_y$$

$$= 2 \|f\|_2^2$$

The second last equality is left as an exercise.

Proposition 4.3.2. Let $f \in \mathcal{H}^2(V)$. Then,

$$\|\Delta f\|_2^2 \le 2\mathcal{E}(f, f)$$

Proof.

$$\|\Delta f\|_{2}^{2} = \sum_{x \in V} (\Delta f(x))^{2} \mu_{x}$$

$$= \sum_{x \in V} \left[\frac{1}{\mu_{x}} \sum_{y \in V} (f(y) - f(x))^{2} \mu_{xy} \right]^{2} \mu_{x}$$

$$= \sum_{x \in V} \frac{1}{\mu_{x}} \left[\sum_{y \in V} (f(x) - f(y))^{2} \mu_{xy} \right]^{2}$$

$$\stackrel{Ex}{\leq} \sum_{x \in V} \frac{1}{\mu_{x}} \left[\sum_{y \in V} (f(x) - f(y))^{2} \mu_{xy} \right] \left[\sum_{y \in V} \mu_{xy} \right]$$

$$= 2\mathcal{E}(f, f)$$

The second last inequality is an exercise and can be shown using Cauchy-Schwarz inequality. \Box

Proof of Discrete Green's Theorem (4.3.1).

$$\begin{split} \langle \Delta f, \ g \rangle &= \sum_{x \in V} \Delta f(x) g(x) \mu_x \\ &= \sum_{x \in V} \frac{1}{\mu_x} \sum_{y \in V} (f(y) - f(x)) \mu_{xy} g(x) \mu_x \\ &= - \sum_{x \in V} \sum_{y \in V} (f(x) - f(y)) \mu_{xy} g(x) \end{split}$$

$$\mathcal{E}(f,g) = \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (f(x) - f(y))(g(x) - g(y))\mu_{xy}$$

$$= \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (f(x) - f(y))g(x)\mu_{xy} - \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (f(x) - f(y))g(y)\mu_{xy}$$

$$= -\frac{1}{2} \langle \Delta f, g \rangle - \frac{1}{2} \sum_{y \in V} \sum_{x \in V} (f(x) - f(y))g(y)\mu_{xy}$$

$$\stackrel{*}{=} -\frac{1}{2} \langle \Delta f, g \rangle - \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (f(y) - f(x))g(x)\mu_{yx}$$

$$= -\frac{1}{2} \langle \Delta f, g \rangle + \frac{1}{2} \sum_{x \in V} \sum_{y \in V} (f(x) - f(y))g(x)\mu_{xy}$$

$$= -\frac{1}{2} \langle \Delta f, g \rangle - \frac{1}{2} \langle \Delta f, g \rangle$$

$$= -\langle \Delta f, g \rangle$$

where * is obtained by flipping the labels of x and y.

Example.

Let $V = \mathbb{N}$ and μ be the usual weights.

Define $f, g : \mathbb{N} \to \mathbb{R}$ such that

$$f(n) := \sum_{i=1}^{n} \frac{(-1)^{i}}{i}$$
$$g(n) := 1$$

Then,

$$\mathcal{E}(f,f) = \frac{1}{2} \left[\sum_{k \ge 1} (f(k+1) - f(k))^2 + \sum_{k \ge 1} (f(k-1) - f(k))^2 \right]$$
$$\le \sum_{k \ge 2} \frac{1}{k^2} < \infty$$

$$\mathcal{E}(g,g) = 0$$

$$\mathcal{E}(f,g) = 0$$

$$\Delta f(n) = \frac{1}{2} [f(n+1) + f(n-1) - 2f(n)]$$

$$= \frac{(-1)^{n+1}}{2} \frac{2n+1}{n(n+1)}$$

$$\implies \langle \Delta f, g \rangle = \frac{3}{4} + \sum_{n=2}^{\infty} \frac{(-1)^{n+1}}{2} \frac{2n+1}{n(n+1)} 2$$

$$= \frac{3}{4} - \frac{1}{2} \neq 0$$

which contradicts the Discrete Green's Theorem (4.3.1)