

# ELECTRICAL NETWORKS AND REVERSIBLE MARKOV CHAINS

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The object of this part of the course is to show how the mathematics of finite electrical networks can be used in the study of reversible Markov chains on finite and countable state spaces. The basis of the connection is that *harmonic functions* for reversible Markov chains can be interpreted as *voltages* for electrical networks. Thus, techniques for calculating, approximating, and bounding voltages (especially *shorting* and *cutting*) may be applied to problems involving harmonic functions (and, in particular, hitting probabilities).

## 1. HITTING PROBABILITIES AND HARMONIC FUNCTIONS

### 1.1. Harmonic Functions and Hitting Probabilities.

**Assumption 1.** Assume henceforth that all Markov chains are irreducible and that their associated graphs or digraphs are locally finite.

**Definition 1.** Let  $F$  be a subset of  $\mathcal{V}$ . A function  $h : F \cup \partial_{\text{out}}F \rightarrow \mathbb{R}$  is said to be *harmonic* at a vertex  $x \in F$  if

$$(1) \quad h(x) = \sum_y p(x, y)h(y)$$

It is said to be *harmonic in  $B$*  if it is harmonic at every  $x \in B$ , and is said to be *harmonic* if it is harmonic on the entire state space  $\mathcal{V}$ .

Since the underlying graph is *locally finite* (meaning that no vertex has infinitely many nearest neighbors) the sum in (1) has only finitely many nonzero terms. Observe that any linear combination of harmonic functions is harmonic: this is known as the *Superposition Property*. In particular, the *difference* between two harmonic functions is harmonic. The term “harmonic” is used because the harmonic functions of classical mathematics are those that satisfy the “mean value property”.

The next proposition is called the *Maximum Principle* for harmonic functions.

**Proposition 2.** Let  $B \subset \mathcal{V}$  be a finite set of vertices whose out-boundary  $\partial_{\text{out}}B \neq \emptyset$ . Suppose that  $h : B \cup \partial B \rightarrow \mathbb{R}$  is harmonic in  $B$ . Then  $h$  has no nontrivial local maximum in  $B$ , that is, there is no vertex  $x \in B$  such that

$$\begin{aligned} h(x) &\geq h(y) \quad \text{for all } y \in \mathcal{E}_x^+ \quad \text{and} \\ h(x) &> h(y) \quad \text{for some } y \in \mathcal{E}_x^+. \end{aligned}$$

Furthermore, if  $\partial_{\text{out}}B \neq \emptyset$  then

$$(2) \quad \max_{x \in B} h(x) \leq \max_{x \in \partial B} h(x).$$

If  $B$  is connected then the inequality is strict unless  $h$  is constant on  $B \cup \partial B$ .

*Proof.* The first statement is fairly obvious. If  $h$  is harmonic at  $x$  then the equation (1) holds, so  $h(x)$  is a (weighted) average of the values  $h(y)$ , where  $y$  is an out-neighbor of  $x$ . Hence, if  $h(x) \geq h(y)$  for all such  $y$  then it must be that  $h(x) = h(y)$  for all such  $y$ .

Because  $B$  is finite,  $h$  must achieve a maximum  $m$  at some  $x \in B$ . Pick one such  $x$ . Then  $h(y) = h(x) = m$  for all out-neighbors  $y$  of  $x$ , and similarly  $h(z) = h(y) = m$  for all out-neighbors  $z$  of out-neighbors  $y$ , and so on. Because the state space is connected (by assumption, all Markov chains considered here are irreducible), there must be a path from  $x$  to  $\partial_{out} B$  along which  $h = m$  (unless  $\partial_{out} B = \emptyset$ ). Since this path ends at  $\partial_{out} B$ , it follows that there is at least one  $y \in \partial_{out} B$  such that  $h(y) = m$ . This proves (2). The proof of the last statement is similar and is left as an exercise.  $\square$

**Corollary 3.** (*Uniqueness Theorem*) *If  $f, g$  are harmonic on  $B$ , and if  $f = g$  on  $\partial B$ , then  $f = g$  in  $B$ , provided  $B$  is finite and  $\partial_{out} B \neq \emptyset$ .*

*Proof.* The difference  $h = f - g$  is harmonic in  $B$  and zero on  $\partial B$ , so the Maximum Principle implies that  $h = 0$  in  $B$ .  $\square$

The importance of harmonic functions for us is that they determine the *hitting probabilities* of the Markov chain. For any subset  $B$  of  $\mathcal{V}$ , define  $T = T_B$  to be the time of first exit from  $B$ , i.e.,

$$T = T_B = \inf\{n \geq 0 : X_n \notin B\}.$$

**Proposition 4.** *Let  $B \subset \mathcal{V}$  be such that  $P^x\{T_B < \infty\} = 1$  for every  $x \in B$ . Suppose that  $\partial_{out} B = F \cup G$ , where  $F$  and  $G$  are disjoint sets. Define*

$$h(x) = P^x\{X_T \in G\} \quad \text{for } x \in B \cup \partial_{out} B.$$

*Then  $h$  is harmonic in  $B$  and satisfies the boundary conditions  $h \equiv 0$  on  $F$  and  $h \equiv 1$  on  $G$ . Moreover, it is the unique bounded harmonic function satisfying these boundary conditions.*

*Proof.* That  $h$  is harmonic in  $B$  is an easy consequence of the Markov property. (Condition on the first step  $X_1$ .) If  $B \cup \partial_{out} B$  is finite, then there can only be one harmonic function satisfying the boundary conditions  $h \equiv 0$  on  $F$  and  $h \equiv 1$  on  $G$ , by the Maximum Principle (Proposition 2) and the Superposition Principle. If  $B$  is infinite then uniqueness follows from the following proposition.  $\square$

**Proposition 5.** *If  $h : \mathcal{V} \rightarrow \mathbb{R}$  is a bounded (nonnegative) function that is harmonic in  $B$ , then  $h(X_{n \wedge T})$  is a bounded (nonnegative) martingale under any  $P^x$ .*

The proof is an elementary exercise in the use of the Markov property and the definition of a martingale (exercise). The martingale property provides another easy proof of uniqueness in Proposition 4. Suppose that there were *two* bounded harmonic functions both satisfying the same boundary conditions. Then their difference  $h$  would be a harmonic function in  $B$  such that  $h \equiv 0$  in  $\partial B$ , and since  $T = T_B < \infty$  a.s., it would follow that  $h(X_T) = 0$  a.s. But then, by the optional sampling theorem for bounded martingales,  $h(x) = E^x h(X_0) = E^x h(X_{T \wedge n})$  for all  $n \geq 0$  and all  $x \in B$ . Consequently, by the DCT,

$$h(x) = \lim_{n \rightarrow \infty} E^x h(X_{T \wedge n}) = E^x h(X_T) = 0,$$

proving that the two harmonic functions were actually identical.

**1.2. Harmonic Functions: Examples.** Proposition 4 provides a road map for solving “Gambler’s Ruin” problems for Markov chains. It is, of course, effective only when harmonic functions can be identified, but there are many examples where this can be done.

**Example 6.** Consider simple random walk  $S_n$  on the integers  $\mathbb{Z}$ . It is trivial to check that *any linear function*  $h(x) = ax + b$  is harmonic. Let  $B = [m] = \{1, 2, \dots, m\}$ , and define  $T = T_B$  to be the time of first exit from  $B$ , as above. If the initial point of the Markov chain is in  $B \cup \partial B$  then with  $P^x$ –probability 1 the exit point  $S_T = 0$  or  $m + 1$ . To find the probability  $h(x)$  that  $S_T = x$  we must find a harmonic function such that  $h(0) = 0$  and  $h(m + 1) = 1$ . This is easy:

$$h(x) = \frac{x}{m+1}.$$

**Example 7.** Let  $X_n$  be  $p - q$  random walk on the integers, with  $p > 1/2 > q > 0$ . It is routine to check that the following are all harmonic functions:

$$h(x) = A + B \left( \frac{q}{p} \right)^x.$$

To solve the Gambler’s Ruin problem here, look for a harmonic function such that  $h(0) = 0$  and  $h(m + 1) = 1$ . The one that works is

$$h(x) = \frac{(q/p)^x - 1}{(q/p)^{m+1} - 1}.$$

## 2. ELECTRICAL NETWORKS

**2.1. Reversible Markov Chains.** Let  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  be a graph<sup>1</sup> or digraph, where  $\mathcal{V}$  is the set of *vertices* (also called *nodes*, or in the context of Markov chains, *states*), and  $\mathcal{E}$  is the set of *edges* (or *directed edges*, if  $\mathcal{G}$  is a digraph). A *Markov chain* on  $\mathcal{G}$  is a Markov chain whose state space is the vertex set  $\mathcal{V}$  of  $\mathcal{G}$ , and whose transition probabilities  $p(x, y)$  satisfy the condition

$$p(x, y) > 0 \quad \text{if and only if } (x, y) \in \mathcal{E}.$$

It is called *reversible* if there is a function  $w : \mathcal{V} \rightarrow (0, \infty)$  such that the following *detailed balance equation* holds for all  $x, y \in \mathcal{V}$ :

$$(3) \quad w_x p(x, y) = w_y p(y, x).$$

Many interesting Markov chains are reversible. The electrical network connection is valid only for reversible Markov chains; hence our interest in them.

### Examples:

- (1) Any Markov chain on a tree.
- (2) Any birth and death chain on the integers.
- (3) Any *homogeneous* nearest neighbor random walk on the  $d$ –dimensional integer lattice  $\mathbb{Z}^d$ . *Homogeneous* means that for any  $x, y, z \in \mathbb{Z}^d$ ,

$$p(x, y) = p(x + z, y + z).$$

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<sup>1</sup>We will also allow  $\mathcal{G}$  to be a *multigraph*, that is, a graph which might have more than one edge connecting a given pair of vertices. Moreover, we will allow  $\mathcal{G}$  to have *self-loops*, that is, edges connecting a vertex to itself.

EXERCISE: Find (or construct) the relevant weight functions  $w$ .

Until further notice, we will assume that all Markov chains are irreducible, i.e., that for any two vertices there is a positive probability path connecting them. This implies that the underlying graph  $\mathcal{G}$  is connected. Notice that if a reversible Markov chain is irreducible then the weight function  $w$  is unique up to multiplication by a scalar (proof by induction on the distance from a distinguished vertex). Before giving examples of reversible Markov chains we will state a basic result due to Kolmogorov that partially explains the terminology and gives a useful tool for verifying that a Markov chain is reversible.

**Proposition 8.** *An irreducible Markov chain with transition probabilities  $p(x, y)$  is reversible iff for every closed path  $x_0, x_1, x_2, \dots, x_n = x_0$ ,*

$$\prod_{i=0}^{n-1} p(x_i, x_{i+1}) = \prod_{i=0}^{n-1} p(x_{i+1}, x_i).$$

The proof is easy, and left as an exercise. At first sight it looks important, but in fact we will have no further use for it. The following is also true:

**Proposition 9.** *Suppose that the detailed balance equations hold with weight function  $w$ . If  $\sum_{x \in \mathcal{V}} w_x < \infty$  then the Markov chain is positive recurrent, and  $\pi_x = w_x / \sum_{x \in \mathcal{V}} w_x$  is a stationary probability distribution. Moreover, the detailed balance equations hold with  $w$  replaced by  $\pi$ .*

The proof is a one-liner (at any rate, if the line is sufficiently long).

**2.2. Electrical Networks and Reversible Markov Chains.** An *electrical network* is a finite graph (or multigraph, or multigraph with self-loops)  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  together with an assignment of *resistances*  $R_e = R(e)$  to the edges. Given a resistance function  $R$ , one defines the *conductance* of any edge to be the reciprocal of its resistance:

$$C_e = C(e) = 1/R(e).$$

Say that an electrical network is *connected* if the underlying graph  $\mathcal{G}$  is connected and  $C_e > 0$  for every edge  $e$ .

STANDING ASSUMPTION: Unless otherwise specified, all electrical networks considered are connected and locally finite. (A graph is *locally finite* if for each vertex  $x$  there are only finitely many edges incident to  $x$ .)

There is a natural one-to-one correspondence between electrical networks without multiple edges and reversible Markov chains (either in discrete or continuous time). Consider a finite connected graph  $\mathcal{G}$  with no multiple edges, and let  $\{C_e\}_{e \in \mathcal{E}}$  be a set of *strictly positive* conductances. For each pair  $x, y$  of vertices (not necessarily distinct!) such that there is an edge  $e = xy$  connecting  $x$  and  $y$ , define

$$(4) \quad p(x, y) = \frac{C_{xy}}{C_x}$$

where for each vertex  $x$  the normalizing constant  $C_x$  is the sum of the conductances  $C_e$  over all edges incident to  $x$ . Clearly, any function  $p(x, y)$  so defined is a transition probability kernel,

and because the conductances are positive, the Markov chain with these transition probabilities is *irreducible*. Furthermore, because conductances are non-directional (that is,  $C_{xy} = C_{yx}$ ), the detailed balance equations hold with weight function  $w_x = C_x$ :

$$C_x p_{x,y} = C_{xy} = C_y p_{y,x}.$$

Conversely, given a transition probability kernel satisfying the detailed balance equations (3), one may unambiguously define conductances by

$$(5) \quad C_{xy} = w(x)p(x,y) = w(y)p(y,x).$$

(For electrical networks on multigraphs, or multigraphs with loops, it is also possible to define in a natural way a corresponding reversible Markov chain, but we won't have any need for this: we will only use electrical networks on multigraphs as technical aids for "solving" electrical networks on graphs.)

**Important Observation:** The definition of a harmonic function may be given in terms of the conductances  $C_e$  as follows:  $h$  is harmonic at  $x \in \mathcal{V}$  iff

$$(6) \quad \sum_{e \in \mathcal{E}_x} C(e)(h(x) - h(y)) = 0,$$

where  $x, y$  are the endpoints of  $e$ . We will see that harmonic functions have interpretations as *electrical potentials*.

**2.3. Flows in Graphs and Multigraphs.** Let  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  be a graph or multigraph, and let  $\mathcal{E}^\circ$  be the set of self-loops (edges with only one incident vertex). Each edge  $e \notin \mathcal{E}^\circ$  has two incident vertices (endpoints)  $x, y$ , and so may be *oriented* (or *directed*) in two ways: from  $x$  to  $y$ , or from  $y$  to  $x$ . Write  $e_{xy} = e(x, y)$  and  $e_{yx} = e(y, x)$  for these two orientations. For each vertex  $x$ , define

$$\begin{aligned} \mathcal{E}_x^+ &= \mathcal{E}^+(x) = \{e(x, y) : e \in \mathcal{E}_x\} & \text{and} \\ \mathcal{E}_x^- &= \mathcal{E}^-(x) = \{e(y, x) : e \in \mathcal{E}_x\} \end{aligned}$$

to be the sets of oriented edges *leaving* and *entering* vertex  $x$ , respectively. Observe that each of the sets  $\mathcal{E}^+(x)$  and  $\mathcal{E}^-(x)$  is in one-to-one correspondence with the set  $\mathcal{E}(x)$  of edges incident to  $x$ . Define  $\mathcal{E}^*$  to be the set of *all* oriented edges of  $\mathcal{G}$ . Observe that  $(\mathcal{V}, \mathcal{E}^*)$  is a directed graph with the special property that there is a natural pairing of directed edges, namely,  $e(x, y) \leftrightarrow e(y, x)$ .

**Flow:** A *flow* in  $\mathcal{G}$  is a function  $J : \mathcal{E}^* \rightarrow \mathbb{R}$  satisfying the following two properties:

(a) The function  $J$  is *antisymmetric*, i.e., for any edge  $e$  with distinct endpoints  $x, y$ ,

$$(7) \quad J_{e_{xy}} = -J_{e_{yx}}.$$

(b) For every vertex  $x$ ,

$$(8) \quad \sum_{e \in \mathcal{E}^+(x)} J(e(x, y)) = \sum_{e \in \mathcal{E}^-(x)} J(e(y, x)) = 0.$$

If equation (8) holds at all vertices  $x$  *except* vertices  $a$  and  $b$ , then  $J$  is called a flow with *source*  $a$  (or  $b$ ) and *sink*  $b$  (or  $a$ ) depending on which case of the following lemma obtains.

**Lemma 10.** *Suppose that  $J$  is a function that satisfies (7) for every edge  $e$ , and satisfies (8) at every vertex  $x$  except vertices  $a$  and  $b$ . Then either*

$$(9) \quad J(a+) \triangleq \sum_{e \in \mathcal{E}(a)} J(e(a, y)) = J(b-) \triangleq \sum_{e \in \mathcal{E}(b)} J(e(y, b)) > 0 \quad \text{or}$$

$$(10) \quad J(a+) = \sum_{e \in \mathcal{E}(a)} J(e(a, y)) = J(b-) = \sum_{e \in \mathcal{E}(b)} J(e(y, b)) < 0.$$

*Proof.* Exercise. □

**Acyclicity:** A *cycle* in  $\mathcal{G}$  is a finite sequence  $e(x_0 x_1), e(x_1 x_2), \dots, e(x_{n-1} x_n)$  of oriented edges such that (i) for each  $i$ , the *tail* of  $e(x_i x_{i+1})$  is the *head* of  $e(x_{i-1} x_i)$ ; and (ii)  $x_0 = x_n$ . An antisymmetric function  $J : \mathcal{E}^* \rightarrow \mathbb{R}$  is called *acyclic* if it sums to zero around any directed cycle in  $\mathcal{G}$ , i.e., for every cycle,

$$(11) \quad \sum_{i=0}^{n-1} J(e(x_i x_{i+1})) = 0.$$

**Lemma 11.** *Every acyclic antisymmetric function is a gradient, i.e., if  $J$  is antisymmetric and acyclic then there exists a function  $\varphi : \mathcal{V} \rightarrow \mathbb{R}$  on the set of vertices (sometimes called the potential function) such that for every oriented edge  $e(x, y)$  pointing from vertex  $x$  to vertex  $y$ ,*

$$(12) \quad J(e(x, y)) = \varphi(y) - \varphi(x).$$

*Proof.* It is enough to consider the case of a *connected* graph or multigraph. Fix a vertex  $x_* = x_0$ , and define  $\varphi(x_*) = 0$ . For any other vertex  $y$  there is a directed path from  $x_*$  to  $y$ , through vertices  $x_0, x_1, \dots, x_n = y$ . Define

$$\varphi(y) = \sum_{i=0}^{n-1} J(e(x_i x_{i+1})).$$

That this is a valid definition follows from the hypothesis (11), since this guarantees that for any two directed paths from  $x_*$  to  $y$ , the sums are equal. The desired relation (12) follows by choosing the right paths. □

**2.4. Electrical Current: Ohm's and Kirkhoff's Laws.** If the two terminals of a 1-volt battery are attached to vertices  $a$  and  $b$  of an electrical network, then electrical current will flow through the edges (resistors) of the network. The laws governing this current flow are due to Ohm and Kirkhoff; they are experimentally observable facts. According to these laws, there are two functions

$$I : \mathcal{E}^* \longrightarrow \mathbb{R} \quad \text{and} \\ \Psi : \mathcal{E}^* \longrightarrow \mathbb{R},$$

both satisfying (7), that measure the *potential difference* and the *current flow* across directed edges. Assume that the positive terminal of the battery is attached to  $a$  and the negative terminal to  $b$ .

**Ohm's Law:** *For every edge  $e$  with endpoints  $x, y$ ,*

$$(13) \quad \Psi(e(x y)) = I(e(x y))R(e).$$

**Kirchhoff's Potential Law:** *The potential difference function  $\Psi$  is acyclic.*

**Kirchhoff's Current Law:** *The current  $I$  is a flow with source  $a$  and sink  $b$ .*

**Proposition 12.** *There is a unique function  $\varphi : \mathcal{V} \rightarrow [0, 1]$ , called the voltage or electrical potential, such that*

- (a)  $\varphi(a) = 1$  and  $\varphi(b) = 0$ .
- (b)  $\varphi$  is harmonic at every vertex  $x \neq a, b$ .
- (c) For every edge  $e$  with endpoints  $x, y$ ,

$$(14) \quad \Psi(e(x, y)) = \varphi(x) - \varphi(y).$$

*Proof.* Since the potential difference function  $\Psi$  is acyclic, by Kirchhoff's first law, Lemma 11 implies that it is the gradient of a real-valued function  $\varphi$  on the vertices. This function may be so chosen that its value at  $b$  is 0. Since the attached battery is a 1-volt battery, it follows that  $\varphi(a) = 1$ . That  $\varphi$  is harmonic except at  $a$  and  $b$  follows from Ohm's Law and Kirchhoff's second law, because the definition of a flow implies, for every vertex  $x \neq a, b$ , that

$$\begin{aligned} \sum_{e \in \mathcal{E}(x)} I(e(x, y)) &= 0 && \implies \\ \sum_{e \in \mathcal{E}(x)} \Psi(e(x, y))/R(e) &= 0 && \implies \\ \sum_{e \in \mathcal{E}(x)} \Psi(e(x, y))C(e) &= 0 && \implies \\ \sum_{e \in \mathcal{E}(x)} (\varphi(x) - \varphi(y))C(e) &= 0, \end{aligned}$$

and this implies that  $\varphi$  is harmonic at  $x$ , by equation (6). Uniqueness now follows from the Maximum Principle for harmonic functions. That  $\varphi$  can only take values between 0 and 1 follows from the maximum principle.  $\square$

**Corollary 13.** *The voltage  $\varphi(x)$  at any vertex  $x$  is the probability that the Markov chain on  $\mathcal{G}$  with transition probabilities (4) and initial point  $X_0 = x$  will visit  $a$  before  $b$ .*

**2.5. Effective Resistance and Escape Probabilities.** Consider the electrical network on a *connected* graph  $\mathcal{G}$  with resistance function  $R$  and conductance function  $C = 1/R$ . Denote by  $X_0, X_1, X_2, \dots$  the Markov chain with transition probabilities  $p(x, y)$  given by (4), and such that under the probability measure  $P^x$  the initial state  $X_0 = x$  a.s. For any two vertices  $a \neq b$ , define the escape probability

$$(15) \quad p_{\text{escape}}(a; b) = P^a \{\text{no return to } a \text{ before first visit to } b\}.$$

Define the *effective resistance*  $R_{\text{eff}} = R_{\text{eff}}(a, b)$  between the vertices  $a$  and  $b$  by

$$(16) \quad R_{\text{eff}} = \frac{1}{I(b-)} = \frac{1}{I(a+)}$$

where  $I(a+)$  and  $I(b-)$  denote the net current flow out of  $a$  and into  $b$  (recall that these are equal – see Lemma 10). This equation may be thought of as an extension of Ohm's Law, because this is what the resistance would have to be if the network were to be replaced by a single resistor

between  $a$  and  $b$  in such a way that the total current flow and the voltage differential were the same as for the original network.

The next theorem makes clear the importance of the notion of effective resistance. Recall that  $C_x = \sum_{e \in \mathcal{E}_x} C(e)$  is the total conductance out of  $x$ .

**Theorem 14.**  $p_{\text{escape}}(b; a) = 1/(C_b R_{\text{eff}})$ .

*Proof.* Suppose that the first step of the Markov chain is to vertex  $x$ . Then by our earlier characterization of harmonic functions as hitting probabilities, the conditional probability of escape to  $b$  before return to  $a$  is  $\varphi(x)$ . Consequently,

$$\begin{aligned} p_{\text{escape}}(b; a) &= \sum_x p(b, x) \varphi_x \\ &= \sum_x p(b, x) (\varphi_x - \varphi_b) \\ &= \sum_x C_{bx} (\varphi_x - \varphi_b) / C_b \\ &= \sum_x I_{bx} / C_b \\ &= I_b / C_b \\ &= 1 / C_b R_{\text{eff}}. \end{aligned}$$

□

The preceding theorem is the keystone of the theory to follow. We will use it for two purposes: (1) explicit computations of escape probabilities; and (2) transience/recurrence theorems. First, however, we will develop some further characterizations of the electrical potential and the electrical current flow.

**2.6. Solving Electrical Networks.** For our purposes, “solving” an electrical network means finding the effective resistance between two vertices  $a$  and  $b$ . (Our interest in computing effective resistances stems from their usefulness in computing escape probabilities.) The basic strategy is to successively replace the network by simpler and simpler networks without changing the effective resistance between  $a$  and  $b$ . Here it is important to allow multiple edges between 2 vertices (thus, in this section  $\mathcal{G}$  is in general a “multigraph”).

Say that an electrical network  $(\mathcal{G}, R)$  may be *reduced* to the network  $(\mathcal{G}', R')$  if  $a$  and  $b$  are vertices in both networks and the effective resistance between  $a$  and  $b$  is the same for both  $(\mathcal{G}, R)$  and  $(\mathcal{G}', R')$ . We will show that there are 3 different kinds of elementary reductions that can be made: these are known as the “Series Law”, the “Parallel Law”, and the “Shorting Law”. Informally, these may be stated as follows:

**Series Law:** Resistances in series add.

**Parallel Law:** Conductances in parallel add.

**Shorting Law:** Vertices at the same voltage may be shorted.

More precisely,

**Proposition 15.** (*Series Law*) Suppose that vertices  $x_0, x_1, x_2, \dots, x_n$  appear in series, i.e., there is a single resistor of resistance  $R_i$  connecting  $x_i$  to  $x_{i+1}$  for each  $i = 1, 2, \dots, n$ , and there are no



other edges emanating from any of the vertices  $x_1, x_2, \dots, x_{n-1}$ . Assume that the source  $a$  and the sink  $b$  are not in the series  $x_0, x_1, x_2, \dots, x_n$ . Then the network may be reduced to the network in which the vertices  $x_1, x_2, \dots, x_{n-1}$  and the edges emanating from them are removed and replaced by a single resistor between  $x_0$  and  $x_n$  with resistance  $R_1 + R_2 + \dots + R_n$ .

**Proposition 16.** (Parallel Law) Suppose that vertices  $x, y$  are connected by edges  $\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n$  with resistances  $R_1, R_2, \dots, R_n$ . Then the network may be reduced to the network in which the edges  $\varepsilon_1, \varepsilon_2, \dots, \varepsilon_n$  are replaced by a single edge of resistance  $(R_1^{-1} + R_2^{-1} + \dots + R_n^{-1})^{-1}$  (i.e., an edge whose conductance is the sum of the conductances of the edges replaced).

**Proposition 17.** (Shorting Law) If two vertices  $x, y$  have the same voltage, then the network may be reduced to the network in which  $x, y$  are replaced by a single vertex  $z$  and the edges with a vertex at  $x$  or  $y$  are rerouted to the vertex  $z$ .

The reason this is called “shorting” is that in effect one is introducing a new wire of *infinite* conductivity (zero resistance) between  $x$  and  $y$ .

The proofs of all three laws are easy, and should be done as EXERCISES.

**2.7. Examples.** Three simple examples of the above laws follow.

**Example 18. Simple Nearest Neighbor Random Walk on the Integers** Consider simple random walk on the set of integers  $\{0, 1, 2, \dots, m\}$ , with reflection at the endpoints  $0, m$ . This is the reversible Markov chain corresponding to the electrical network with unit resistances between  $i$  and  $i + 1$  for all  $i = 0, 1, \dots, m - 1$ . The series law implies that the effective resistance between  $0$  and  $m$  is  $m$ . It follows that the probability of escape from  $0$  to  $m$  is  $1/m$ . Note that as  $m \rightarrow \infty$  the escape probability drops to  $0$ ; this implies that the simple nearest neighbor random walk on the integers is recurrent.

**Example 19. p-q Random Walk on the Integers** Consider the network on the graph whose vertex set is the set of positive integers and whose edges are between nearest neighbors. The conductance function is as follows:

$$\begin{aligned} C_{0,1} &= 1; \\ C_{x,x+1} &= \left(\frac{p}{q}\right) C_{x-1,x} \\ &= \left(\frac{p}{q}\right)^x. \end{aligned}$$

The reversible Markov chain corresponding to this conductance function is the p-q random walk on the positive integers. The effective resistance between  $0$  and  $b$  can be found using the series law: the series between  $0$  and  $b$  can be replaced by a single edge from  $0$  to  $b$  of resistance

$$\sum_{x=0}^{b-1} \left(\frac{q}{p}\right)^x = \frac{1 - (q/p)^b}{1 - (q/p)}.$$

This is obviously the effective resistance. It follows that the escape probability from  $0$  to  $b$  is  $1$  over the effective resistance, which is

$$\frac{1 - (q/p)}{1 - (q/p)^b}.$$

**Example 20. Simple Random Walk on the Cube** Consider the simple nearest neighbor random walk on the vertices of the cube  $\mathbb{Z}_2^3$ . This is the reversible Markov chain corresponding to the electrical network whose vertices and edges are the vertices and edges of the cube, and for which the conductances of the edges are all 1. We will calculate the effective resistance and the escape probability for two neighboring vertices,  $a = 000$  and  $b = 001$ . Symmetry implies that the voltage function satisfies

$$\begin{aligned} h(010) &= h(100); \\ h(011) &= h(101). \end{aligned}$$

The shorting law implies that the network may be reduced by shorting these 2 triples of vertices. This leaves a network with multiple edges that may be successively reduced by applying the parallel and series laws in sequence. One finds (I think) that the effective resistance between  $a$  and  $b$  is  $4/3$ , so the escape probability is  $3/4$ .

### 3. VARIATIONAL PRINCIPLES

Throughout this section, let  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  be a finite graph (no multiple edges or loops) equipped with resistance function  $R$  and conductance function  $C$ . Let  $I$  be the electrical current function and  $\varphi$  the voltage function when a  $v$ -volt battery is attached to vertices  $a$  and  $b$ . The goal of this section is to show that  $I$  and  $\varphi$  satisfy simple *variational principles*.

Let  $\ell^2(\mathcal{V})$  and  $\ell^2(\mathcal{E}^*)$  denote the vector spaces of real-valued functions on the sets of vertices and oriented edges, respectively, equipped with the inner products

$$(17) \quad \langle f, g \rangle = \langle f, g \rangle_{\ell^2(\mathcal{V})} = \sum_{x \in \mathcal{V}} f(x)g(x) \quad \text{and}$$

$$(18) \quad \langle f, g \rangle = \langle f, g \rangle_{\ell^2(\mathcal{E}^*)} = \sum_{e(x,y) \in \mathcal{E}^*} f(e(x,y))g(e(x,y)).$$

Let

$$(19) \quad \ell_{\pm}^2(\mathcal{E}^*) = \{\text{antisymmetric } f \in \ell^2(\mathcal{E}^*)\}$$

be the subspace of  $\ell^2(\mathcal{E}^*)$  consisting of all *antisymmetric* functions (functions that satisfy equation (7)). There is a natural linear transformation from  $\ell^2(\mathcal{V})$  to  $\ell_{\pm}^2(\mathcal{E}^*)$ , called the *gradient* map  $\nabla$ , defined by

$$(20) \quad \nabla f(e_{xy}) = f(y) - f(x).$$

**Dirichlet Forms:** The Dirichlet form associated with the conductance function  $C$  is the quadratic form on  $\ell_{\pm}^2(\mathcal{E}^*)$  defined by

$$(21) \quad \mathcal{D}(f, g) = \frac{1}{2} \sum_{e(x,y) \in \mathcal{E}^*} f(e_{xy})g(e_{xy})R(e).$$

Observe that each edge  $e \in \mathcal{E}$  is represented twice in this sum, once for each orientation; and the two terms of the sum are equal, since both  $f$  and  $g$  are antisymmetric. (This explains the

factor of 1/2.) There is an analogous quadratic form defined on  $\ell^2(\mathcal{V})$ , which we will call the *dual Dirichlet form*, by

$$(22) \quad \begin{aligned} \mathcal{D}^*(f, g) &= \mathcal{D}(C\nabla f, C\nabla g) \\ &= \frac{1}{2} \sum_{e(xy) \in \mathcal{E}^*} \nabla f(e(xy)) \nabla g(e(xy)) C(e). \end{aligned}$$

**Theorem 21.** (*Dirichlet's Principle, First Form*) Let  $\varphi$  be the electrical potential when a battery of voltage 1 is attached to the vertices  $a$  and  $b$ , i.e.,  $\varphi$  is unique bounded function on  $\mathcal{V}$  that is harmonic on  $\mathcal{V} - \{a, b\}$  and satisfies the boundary conditions  $\varphi(a) = 1$  and  $\varphi(b) = 0$ . Then  $\varphi$  minimizes the Dirichlet form  $\mathcal{D}^*(h, h)$  within the class of functions  $h \in \ell^2(\mathcal{V})$  satisfying the boundary conditions  $h(a) = 1$  and  $h(b) = 0$ .

**Remark 1.** Note: The uniqueness of the minimizing function  $\varphi$  is part of the assertion.

**Theorem 22.** (*Dirichlet's Principle, Second Form*) Let  $I$  be the electrical current flow when a  $v$ -volt battery is hooked up at  $a$  and  $b$ . Then among all flows  $J$  on the graph  $\mathcal{G}$  with source  $a$  and sink  $b$  and satisfying the "total flow" constraint  $J(a+) = I(a+)$ , the flow  $I$  minimizes the Dirichlet form  $\mathcal{D}(J, J)$ . Moreover,

$$(23) \quad \mathcal{D}^*(\varphi, \varphi) = \mathcal{D}(I, I).$$

The proofs will require the following:

**Proposition 23.** (*Conservation Law*) Let  $w : \mathcal{V} \rightarrow \mathbb{R}$  be any real-valued function on the set of vertices, and let  $J$  be a flow on  $\mathcal{G}$  with source  $a$  and sink  $b$ . Then

$$(w_a - w_b)J(a+) = \frac{1}{2} \sum_{e(xy) \in \mathcal{E}^*} (w_x - w_y)J(e(xy)).$$

*Proof.* This is a simple computation: just group terms in the sum according to the initial point of the edge. In detail,

$$\begin{aligned} \frac{1}{2} \sum_{\mathcal{E}^*} (w_x - w_y)J(e(xy)) &= \sum_{\mathcal{E}^*} w_x J(e(xy)) \\ &= \sum_{x \in \mathcal{V}} \sum_{e \in \mathcal{E}(x)} w_x J(e(xy)) \\ &= \sum_{\mathcal{E}(a)} w_a J(e(a y)) + \sum_{\mathcal{E}(b)} w_b J(e(b y)) \\ &= w_a J(a+) + w_b J(b+) \\ &= (w_a - w_b)J(a+). \end{aligned}$$

□

We will only use this as a technical lemma; however, it is an interesting result in its own right in graph theory. It has the following interpretation: think of the flow as representing a flow of some "commodity" in an economic network. At each node  $x$  the value of (a unit amount of) the commodity is determined by the function  $w_x$ . Thus, for each directed edge  $e(xy)$  the quantity  $(w_y - w_x)J(e_{xy})$  is the "value added" to the amount  $J(e_{xy})$  of the commodity that passes

through the edge. Similarly,  $(w_a - w_b)J(a+)$  represents the change in value of the amount  $J(a+)$  of the commodity that enters the network at  $a$  by the time it leaves the network from  $b$ . This should make the conservation law intuitively clear.

*Proof of Theorem 21.* There is clearly a function  $h$  that minimizes the dual Dirichlet form  $\mathcal{D}^*(h, h)$  subject to the constraints  $h(a) = 1, h(b) = 0$ , since this is a finite-dimensional minimization problem. We will argue that this function is harmonic.

Recall that  $C_x$  is the sum of the conductances leading out of  $x$ . Note that  $C_x > 0$  because of our standing assumption that the electrical network is connected. For any directed edge  $e = e(xy) \in \mathcal{E}^+(x)$ , let  $p^x(e) = C(e)/C_x$ ; this is a probability distribution on the edges leading out of  $x$ . The condition that  $h$  is harmonic at  $x$  is equivalent to the statement that

$$h(x) = \sum_{e(xy) \in \mathcal{E}_x^+} p^x(e)h(y).$$

Fix a vertex  $x \neq a, b$ , and consider those terms in the sum defining the dual Dirichlet form  $\mathcal{D}^*(h, h)$  that involve  $h(x)$ : they contribute

$$\sum_{\mathcal{E}_x} (h(x) - h(y))^2 C(e) = C_x \sum_{\mathcal{E}_x} (h(x) - h(y))^2 p^x(e).$$

Here  $y$  denotes the other endpoint of  $e$ . For fixed values of  $h(y)$ ,  $y \neq x$ , the value of  $h(x)$  that minimizes this is

$$h(x) = \sum_{\mathcal{E}_x} p(e)^x h(y).$$

Hence, if  $h$  were *not* harmonic at  $x$ , then it would not minimize the energy  $\mathcal{D}(h)$  (this argument does not apply at  $x = a, b$ , though, because at those points  $h$  must satisfy the boundary conditions). Thus,  $h$  minimizes the dual Dirichlet form only if it is harmonic. But there is only one such harmonic function, by the Maximum Principle, and it is the electrical potential  $\varphi$ .  $\square$

*Proof of Theorem 22.* Let  $J$  be a flow on  $\mathcal{G}$  with source  $a$  and sink  $b$  such that  $J(a+) = I(a+)$ , where  $I$  is the electrical current flow. Define a third flow  $H$  by taking the difference of  $I$  and  $J$ : note that this flow has no source and no sink. Write  $J = I + H$  and expand the square in  $\Delta^*(J, J)$  to get

$$\begin{aligned} 2\mathcal{D}(J, J) &= \sum_{\mathcal{E}^*} J(e_{xy})^2 R_e \\ &= \sum_{\mathcal{E}^*} (I(e_{xy}) + H(e_{xy}))^2 R_e \\ &= \sum_{\mathcal{E}^*} I(e_{xy})^2 R_e + \sum_{\mathcal{E}^*} H(e_{xy})^2 R_e + \sum_{\mathcal{E}^*} I(e_{xy})H(e_{xy})R_e. \end{aligned}$$

To prove that  $I$  is the minimizer of the energy, it obviously suffices to show that the third sum is 0. By Ohm's Law,  $I(e_{xy})R_e = \varphi(x) - \varphi(y)$ ; thus, by the Conservation Law for flows proved earlier

in this section,

$$\begin{aligned}\sum_{\mathcal{E}^*} H(e_{xy}) I(e_{xy}) R_e &= \sum_{\mathcal{E}^*} H(e_{xy}) (\varphi(x) - \varphi(y)) \\ &= (\varphi(a) - \varphi(b)) H(a+) \\ &= 0,\end{aligned}$$

the last equation because  $H$  has no source or sink.

That  $\mathcal{D}(I, I) = \mathcal{D}^*(\varphi, \varphi)$  follows easily from Ohm's Law.  $\square$

The quantity  $\mathcal{D}(I, I)$  is called the “total energy dissipated” in the network. It is what you are billed for by the electric company. The total energy dissipated can be expressed very simply in terms of the effective resistance between  $a$  and  $b$ .

**Proposition 24.**  $\mathcal{D}(I, I) = \mathcal{D}^*(\varphi, \varphi) = I(a+)^2 R_{\text{eff}}(a, b)$ .

*Proof.* That  $\mathcal{D}^*(\varphi, \varphi) = \mathcal{D}(I, I)$  was proved above. By Ohm's Law and the conservation law for flows,

$$\begin{aligned}2\mathcal{D}(I, I) &= \sum_{\mathcal{E}^*} I(e_{xy})^2 R_e \\ &= \sum_{\mathcal{E}^*} I(e_{xy}) (h(x) - h(y)) C_e R_e \\ &= \sum_{\mathcal{E}^*} I(e_{xy}) (h(x) - h(y)) \\ &= 2(h(a) - h(b)) I(a+) \\ &= 2I(a+)^2 R_{\text{eff}}(a, b).\end{aligned}$$

$\square$

**Corollary 25.**  $R_{\text{eff}}(a, b) = \min\{\mathcal{D}(J, J) : J(a+) = J(b-) = 1\}$ .

#### 4. RAYLEIGH'S MONOTONICITY LAW

The variational principles established earlier are important for various of reasons, not least among them a *comparison principle* for electrical networks due to Lord Rayleigh. It essentially states that if some or all of the resistances in an electrical network are increased then the effective resistance between any 2 vertices cannot decrease. We will only consider ordinary finite graphs, i.e., graphs with no multiple edges or loops.

**Theorem 26.** (*Rayleigh's Monotonicity Principle*) Let  $R$  and  $\bar{R}$  be two resistance functions on the same finite graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  satisfying  $R_e \leq \bar{R}_e \forall e \in \mathcal{E}$ . Then for all  $a, b \in \mathcal{V}$ ,

$$R_{\text{eff}}(a; b) \leq \bar{R}_{\text{eff}}(a; b).$$

*Proof.* Instead of fixing the batteries so that the voltage differential between vertices  $a$  and  $b$  is 1, we choose voltages  $v, \bar{v} > 0$  so that the electrical current flows  $I$  and  $\bar{I}$  satisfy  $I(a+) = \bar{I}(a+) = 1$ . Note that multiplying the voltage by a scalar does not affect the validity of either variational principle. By the second variational principle, the current flows  $I$  and  $\bar{I}$  minimize the

energies  $\mathcal{D}(J, J)$  and  $\overline{\mathcal{D}}(J, J)$  respectively, among all flows with source  $a$ , sink  $b$ , and satisfying  $J(a+) = 1$ . Recall that

$$\begin{aligned} 2\mathcal{D}(J, J) &= \sum_{e \in \mathcal{E}^*} J(e(xy))^2 R_e \quad \text{and} \\ 2\overline{\mathcal{D}}(J, J) &= \sum_{e \in \mathcal{E}^*} J(e(xy))^2 \overline{R}(e). \end{aligned}$$

Therefore, for every flow  $J$  it must be the case that  $\mathcal{D}(J, J) \leq \overline{\mathcal{D}}(J, J)$ . It follows that  $\mathcal{D}(I, I) \leq \overline{\mathcal{D}}(\overline{I}, \overline{I})$ . Hence, since  $\mathcal{D}(I, I) = I_a^2 R_{\text{eff}}(a; b)$ ,

$$\begin{aligned} \mathcal{D}(I, I) &\leq \overline{\mathcal{D}}(\overline{I}, \overline{I}) \implies \\ I_a^2 R_{\text{eff}}(a; b) &\leq \overline{I}_a^2 \overline{R}_{\text{eff}}(a; b) \implies \\ R_{\text{eff}}(a; b) &\leq \overline{R}_{\text{eff}}(a; b). \end{aligned}$$

□

**Corollary 27.** *If some or all the resistances in an electrical network are increased, except those leading out of  $a$ , then  $p_{\text{escape}}(a : b)$  can only decrease.*

This follows immediately from Rayleigh's monotonicity law and the characterization of  $p_{\text{escape}}(a : b)$  in terms of effective resistance proved earlier.

Rayleigh's monotonicity law and its corollary can be effective tools in transience/recurrence problems. We shall give two examples of its use in proving the recurrence of certain reversible Markov Chains.

**Remark 2.** Simple Random Walk on  $\mathbb{Z}^2$  Simple random walk on  $\mathbb{Z}^2$  is the reversible Markov chain on the integer lattice in 2 dimensions with resistance function  $R$  identically 1. Let the vertices of  $\mathbb{Z}^2$  be arranged in concentric squares centered at the origin: thus

$$\begin{aligned} S_0 &= \{(0, 0)\}; \\ S_k &= \{(m_1, m_2) : |m_1| = k \text{ and } |m_2| \leq k \text{ or } |m_2| = k \text{ and } |m_1| \leq k\}. \end{aligned}$$

We are interested in the probability that simple random walk started at the origin escapes  $S_k$  before returning to the origin. To obtain this probability, we modify the network as follows: replace  $\partial S_k$  by a single vertex  $s_*$ ; any edges that connected vertices of  $S_k$  to vertices of  $\partial S_k$  should be rerouted to  $s_*$ ; all vertices of  $\mathbb{Z}^2$  not in  $\cup_0^{k+1} S_j$  should be deleted from the network. The changes described above have no effect on the probability of escape from the origin to the exterior of  $S_k$  before returning to the origin. Note that after the modifications the exterior of  $S_k$  is just the single vertex  $s_*$ ; consequently, the escape probability is given by

$$p_{\text{escape}} = \frac{1}{4R_{\text{eff}}}.$$

Here  $R_{\text{eff}}$  is the effective resistance between the origin and  $s_*$ .

To estimate the effective resistance, we will define a resistance function  $\overline{R}$  that is dominated by  $R$ ; by the corollary to Rayleigh's monotonicity law it will follow that  $\overline{p}_{\text{escape}} \geq p_{\text{escape}}$ . For each  $l = 1, 2, \dots$ , decrease the resistance between any two vertices of  $S_l$  to 0; i.e., "short" the vertices of  $S_l$ . Decreasing the resistance to zero is equivalent to increasing the conductance to  $\infty$ ; by the Dirichlet Principle for the potential function, it must be the case that  $\overline{h}$  is constant

on each of the sets  $S_l$  (otherwise, the energy  $\mathcal{D}^*(\bar{h}, \bar{h})$  would be  $\infty$ ). Now by the shorting law, to solve for the effective resistance between the origin and  $s_*$  we may replace each of the sets  $S_l$  by a single vertex  $s_l$  and reroute all the resistors accordingly.

The resulting network may be reduced using the parallel law, then again by the series law. First note that there are  $8l - 4$  resistors connecting  $s_{l-1}$  to  $s_l$ , so by the parallel law these may be replaced by a single resistor of resistance  $1/(8l - 4)$ . Then by the series law, the series of vertices from  $s_0$  (the origin) to  $s_k$  may be eliminated and the edges replaced by a single resistor of resistance  $\sum_{l=1}^k (8j - 4)^{-1}$  connecting  $s_0$  to  $s_k$ . The result is that

$$\bar{R}_{\text{eff}} = \sum_{l=1}^k (8j - 4)^{-1} \leq R_{\text{eff}}.$$

The last inequality is by Rayleigh's monotonicity law. It follows that as  $k \rightarrow \infty$ , the escape probability for the simple random walk decreases to zero. Consequently, simple random walk on  $\mathbb{Z}^2$  is recurrent.

**Remark 3.** Simple Random Walk on a Subgraph of  $\mathbb{Z}^2$  Let  $\mathcal{G}$  be a subgraph of  $\mathbb{Z}^2$ : i.e., the vertex set and the edge set of  $\mathcal{G}$  are contained in the vertex set and the edge set of the integer lattice  $\mathbb{Z}^2$ . Thus,  $\mathcal{G}$  is obtained by removing edges from the integer lattice. Assume that  $\mathcal{G}$  is connected, and that the origin  $e$  is in the vertex set of  $\mathcal{G}$ . The simple random walk on  $\mathcal{G}$  is the reversible Markov chain corresponding to the electrical network on  $\mathcal{G}$  in which every edge has resistance 1. We will argue that the simple random walk on  $\mathcal{G}$  is recurrent. To show this it suffices to show that the probability of escaping from  $e$  to the exterior of the sphere of radius  $k$  centered at  $e$  before returning to  $e$  decreases to zero as  $k \rightarrow \infty$ . As in the previous example, we estimate this probability by replacing the (possibly) infinite network  $\mathcal{G}$  by the network in which the boundary of  $B_k$  is replaced by a single vertex  $s_k$  and all edges leading from  $B_k$  to  $\partial B_k$  are rerouted to  $s_k$ . Then the escape probability may be written in terms of the effective resistance  $R_{\text{eff}}$  between  $e$  and  $s_k$ . But the Rayleigh monotonicity law implies that the effective resistance is *no smaller* than that for the integer lattice  $\mathbb{Z}^2$ , because  $\mathcal{G}$  is obtained from  $\mathbb{Z}^2$  by increasing the resistances on certain edges (namely, those deleted) to  $\infty$ . We have shown that the simple random walk on  $\mathbb{Z}^2$  is recurrent. Therefore, the effective resistance between  $e$  and  $s_k$  for the entire integer lattice increases to  $\infty$  as  $k \rightarrow \infty$ . It follows (the Rayleigh monotonicity law) that the same is true for the network  $\mathcal{G}$ . This proves that the simple random walk on  $\mathcal{G}$  is recurrent.

## 5. INFINITE NETWORKS: TRANSIENCE AND RECURRENCE

In this section let  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  be an *infinite* graph (no multiple edges) that is *locally finite*, i.e., such that every vertex has at most finitely many edges attached. (This assumption is actually extraneous, but allows us to avoid having to use any theorems from Hilbert space theory.) We assume as usual that the graph is connected. As for finite graphs, let  $\mathcal{E}^*$  denote the set of *oriented* edges, and for any edge  $e$  with endpoints  $x, y$  denote by  $e(x, y) = e_{x, y}$  the corresponding oriented edge with  $x$  and head  $y$ . Let  $R_{x, y} = R_{y, x}$  be resistance attached to the edge  $e$  with endpoints  $x, y$ , and let  $C_{x, y} = 1/R_{x, y}$  be the corresponding conductance. Assume that the resistances are finite and strictly positive. Let  $X_n$  be the reversible Markov chain on  $\mathcal{V}$  corresponding to the conductance function  $C_{x, y}$ .

A flow on  $\mathcal{G}$  with source  $a$  (and no sink) is an antisymmetric real-valued function  $J_{xy} = J(e(xy))$  on the set  $\mathcal{E}^*$  of directed edges that satisfies the flow condition (8) at every vertex  $x$  except  $x = a$ , where

$$(24) \quad J(a+) = \sum_{y \in \mathcal{V}} J_{ay} > 0.$$

The quantity  $J(a+)$  will be called the *size* of the flow. Define the *energy* of the flow  $J$  by

$$(25) \quad \mathcal{D}(J, J) = \frac{1}{2} \sum_{\mathcal{E}^*} J_{xy}^2 R_{xy}.$$

**Theorem 28.** *If there exists a finite energy flow with a single source and no sink on  $\mathcal{G}$  then the Markov chain  $X_n$  is transient.*

**Theorem 29.** *If  $X_n$  is transient, then for every vertex  $a \in \mathcal{V}$  there exists a finite energy flow on  $\mathcal{G}$  with source  $a$ .*

Theorem 28 sometimes gives an effective means of showing that a Markov chain is transient. To show that a Markov chain is transient it is always sufficient to show the existence of a non-constant nonnegative superharmonic function; but in practice it is often easier to build flows than superharmonic functions.

**5.1. Consequences.** Before beginning the proofs of these theorems, let us point out several consequences. First, the 2 theorems together give another approach to “comparisons” of different Markov chains on the vertex set  $\mathcal{V}$ . Specifically,

**Corollary 30.** *Let  $R_{xy}$  and  $\bar{R}_{xy}$  be resistance functions on the graph  $\mathcal{G}$  satisfying  $R \leq \bar{R}$ . Let  $X_n$  and  $\bar{X}_n$  be the corresponding Markov chains. If  $X_n$  is recurrent, then so is  $\bar{X}_n$ .*

*Proof.* If there is a finite energy flow for  $\bar{R}$  then it is also a finite energy flow for  $R$ . □

This corollary may be used to give shorter proofs of some of the theorems proved in the preceding section using the Rayleigh principle.

As stated, Theorem 29 does not appear to be a useful tool for establishing recurrence, because it requires that one show that every flow with source  $a$  and no sink has infinite energy. However, Theorem 29 has a simple corollary that is useful in showing that certain Markov chains are recurrent. Given a distinguished vertex  $a$ , say that a subset  $\Gamma \subset \mathcal{E}$  is a *cutset* for the network if only finitely many vertices can be reached by paths started at  $a$  that contain no edges in  $\Gamma$ .

**Example:** In the integer lattice  $\mathbb{Z}^2$ , with the origin as distinguished vertex, the set  $\Gamma_m$  consisting of all edges connecting vertices  $(x, y)$  with  $\max(x, y) = m$  to vertices  $(x', y')$  with  $\max(x', y') = m + 1$  is a cutset.

**Theorem 31.** *If there is a sequence of nonoverlapping cutsets  $\Gamma_m$  such that*

$$(26) \quad \sum_{m=1}^{\infty} \left( \sum_{e \in \Gamma_m} C(e) \right)^{-1} = \infty$$

*then there is no finite energy flow with source  $a$ , and so the corresponding reversible Markov chain is recurrent.*



*Proof.* For each  $m$ , let  $\mathcal{V}_m$  be the (finite) set of vertices that can be reached from  $a$  via continuous paths with no edges in  $\Gamma_m$ ; and let  $\Gamma_m^+$  be the set of oriented edges  $e(x, y)$  such that  $e \in \Gamma_m$  and such that  $x \in \mathcal{V}_m$  and  $y \notin \mathcal{V}_m$ . Let  $J$  be a unit flow with source  $a$  and no sink (*unit flow* means that  $J(a+) = 1$ ). Then for every  $m \geq 1$ ,

$$\sum_{e(x, y) \in \Gamma_m^+} J(e(x, y)) = 1$$

by an easy calculation (sum over the vertices  $x \in \mathcal{V}_m$ , and use the fact that  $J$  is a unit flow with source  $a$ ). Consequently, by the Cauchy-Schwartz inequality,

$$\begin{aligned} \sum_{e(x, y) \in \Gamma_m^+} J(e(x, y))^2 R(e) \sum_{e(x, y) \in \Gamma_m^+} C(e) &\geq \left( \sum_{e(x, y) \in \Gamma_m^+} J(e(x, y)) R(e)^{1/2} C(e)^{1/2} \right)^2 \\ &= \left( \sum_{e(x, y) \in \Gamma_m^+} J(e(x, y)) \right)^2 = 1, \end{aligned}$$

so

$$\sum_{e(x, y) \in \Gamma_m^+} J(e(x, y))^2 R(e) \geq 1 / \sum_{e(x, y) \in \Gamma_m^+} C(e).$$

It follows that if (26) holds then the energy of the flow  $J$  must be infinite, because

$$\begin{aligned} 2\mathcal{D}(J, J) &= \sum_{\mathcal{E}^*} J(e(x, y))^2 R(e) \\ &\geq \sum_{m=1}^{\infty} \sum_{e(x, y) \in \Gamma_m^+} J(e(x, y))^2 R(e) \\ &\geq \sum_{m=1}^{\infty} \left( \sum_{e(x, y) \in \Gamma_m^+} C(e) \right)^{-1}. \end{aligned}$$

Therefore, by Theorem 29, the corresponding Markov chain is recurrent.  $\square$

**5.2. Examples.** First, we show how the Nash-Williams criterion gives an easy proof of the recurrence of two-dimensional random walks. Then we give two examples showing that the construction of finite-energy flows may be a feasible strategy for establishing transience.

**5.2.1. Nearest neighbor random walk on  $\mathbb{Z}^2$ .** Let  $C$  be any conductance function on the edges of the two-dimensional integer lattice  $\mathbb{Z}^2$  that is bounded away from 0 and  $\infty$ , and let  $X_n$  be the corresponding reversible Markov chain. There is a natural sequence  $\Gamma_m$  of nonoverlapping cutsets, namely,  $\Gamma_m$  is the set of all edges connecting vertices  $(x, y)$  with  $\max(x, y) = m$  to vertices  $(x', y')$  with  $\max(x', y') = m + 1$ . The cardinality of  $\Gamma_m$  is  $8m + 4$ , so

$$\sum_{m=1}^{\infty} \left( \sum_{\Gamma_m} C(e) \right)^{-1} \geq \text{constant} \sum_{m=1}^{\infty} \frac{1}{8m + 4} = \infty.$$

Therefore, by the Nash-Williams criterion, the Markov chain  $X_n$  is recurrent.

5.2.2. *Nearest neighbor random walk on a homogeneous tree.* The homogeneous tree  $\mathcal{T}_d$  of degree  $d$  is the tree in which every vertex has exactly  $d$  edges attached. For the sake of simplicity, take  $d = 3$ . The tree  $\mathcal{T}_3$  may be described as follows: the vertex set  $\mathcal{V}$  is the set of all finite *reduced words* (including the empty word  $e$ ) from the alphabet  $\{a, b, c\}$ , a *reduced word* being defined as a finite sequence of letters in which no letter ( $a, b$ , or  $c$ ) follows itself. Two words  $v, w$  are neighbors iff one may be obtained from the other by attaching a single letter to the end of the other; note that every vertex has exactly 3 neighbors, e.g.,  $aba$  has neighbors  $ab, abab, abac$ . The length of the word  $v$  is just its distance from the empty word  $e$ .

Let  $C_{\cdot}$  be any conductance function on the edges of the tree  $\mathcal{T}_3$  that is bounded away from 0. Then the resistances are bounded above by a constant  $\rho$ . Using this fact, I'll construct a finite energy flow with source  $e$ . For any word  $v \in \mathcal{V}$  of length  $\geq 0$  and any letter  $i \in \{a, b, c\}$  such that  $vi$  is a reduced word, define

$$J_{v,vi} = \frac{1}{3} \cdot \frac{1}{2^{|v|-1}}$$

where  $|v|$  is the word length of  $v$ . It is clear that  $J$  is a flow with source  $e$  (see diagram), and

$$\mathcal{D}(J, J) \leq \rho \sum_{n=0}^{\infty} N_n (3^{-1} 2^{-n+1})^2$$

where  $N_n$  is the number of edges in the tree leading from a word of length  $n$  to a word of length  $n + 1$ . It is easily proved by induction that  $N_n = (3)(2^{n-1})$ . Consequently,

$$\mathcal{D}(J, J) < \infty.$$

5.2.3. *Simple NN RW on  $\mathbb{Z}^3$ .* For each vertex  $x \in \mathbb{Z}^3$  let  $K_x$  be the unit cube with center  $x$  and edges parallel to the coordinate axes. Two vertices  $x, y$  are neighbors iff the cubes  $K_x, K_y$  have a common face  $K_{xy}$ . The resistances are assumed to be bounded above; thus the conductances are bounded away from 0. (For simple NN RW, the conductances are all 1.) I'll construct a finite energy flow. For neighbors  $x, y$ , define

$$J_{xy} = \int \int_{K_{xy}} \nabla \left( \frac{1}{r} \right) \cdot N_{xy} dS$$

where  $r = \text{distance to the origin}$ ,  $dS$  is the surface area element on the square  $K_{xy}$ , and  $N_{xy}$  is the unit normal on  $K_{xy}$  pointing in the direction  $x$  to  $y$ .

That  $J$  is a flow with source at the origin is an immediate consequence of the divergence theorem, because  $\text{div}(\nabla(1/r)) = 0$  everywhere except at the origin (an easy computation). That  $J$  has finite energy may be seen as follows: First, there exists a constant  $B < \infty$  such that  $J_{xy} \leq Br^{-2}$ , where  $r = \min(|x|, |y|)$ . Second, the number of vertices of the lattice  $\mathbb{Z}^3$  at distance approximately  $n$  from the origin is on the order of  $n^2$ . Consequently, for suitable constants  $B', B''$ ,

$$\begin{aligned} \sum_x \sum_y J_{xy}^2 &\leq \sum_{\text{nonzero vertices}} B' r^{-4} \\ &\leq \sum_{n=1}^{\infty} B'' n^2 n^{-4} \\ &< \infty. \end{aligned}$$

This shows that every reversible, nearest neighbor Markov chain on the 3D integer lattice for which the transition probabilities across nearest neighbor bonds are bounded away from 0 is transient.

**5.3. The Dirichlet Principle Revisited.** Let  $a$  be a vertex of the graph  $\mathcal{G}$  and let  $\Omega$  be a *finite* set of vertices containing  $a$ . Define  $p_{\text{escape}}(a; \Omega)$  to be the probability that if started at  $a$  the Markov chain will escape from  $\Omega$  before returning to  $a$ . We will characterize  $p_{\text{escape}}(a; \Omega)$  in terms of flows.

Recall that the boundary  $\partial\Omega$  of the set  $\Omega$  consists of those vertices of  $\mathcal{G}$  not in  $\Omega$  that are nearest neighbors of vertices in  $\Omega$ . Since  $\Omega$  is finite, so is its boundary, because we have assumed that the graph  $\mathcal{G}$  is locally finite. Define a flow on  $\bar{\Omega} = \Omega \cup \partial\Omega$  with source  $a$  and sink  $\partial\Omega$  to be an antisymmetric function  $J$  on the set of oriented edges of  $\mathcal{G}$  that have at least one endpoint in  $\Omega$  such that

$$\begin{aligned} \text{(a)} \quad & \sum_{\mathcal{E}^+(x)} J_{xy} = 0 \quad \forall x \in \Omega - \{a\}; \\ \text{(b)} \quad & \sum_{\mathcal{E}^+(a)} J_{ay} = C. \end{aligned}$$

Define the energy  $\mathcal{D}_\Omega(J, J)$  of the flow  $J$  in the usual way, namely,

$$\mathcal{D}_\Omega(J, J) = \frac{1}{2} \sum J_{xy}^2 R_{xy},$$

where the sum extends over those directed edges of  $\mathcal{G}$  with at least one endpoint in  $\Omega$ ; since there are only finitely many edges involved (recall that  $\Omega$  is finite and because the graph is locally finite), the energy is always finite.

**Proposition 32.** *The escape probability  $p_{\text{escape}}$  is given by*

$$(27) \quad p_{\text{escape}}(a; \Omega) = \frac{1}{C_a \min \mathcal{D}_\Omega(J, J)},$$

where the min is over all flows  $J$  of size 1 on  $\bar{\Omega}$  with source  $a$  and sink  $\partial\Omega$ . Moreover, the flow of size 1 with source  $a$  that minimizes energy is

$$(28) \quad J_{xy} = \frac{C_{xy}(h(x) - h(y))}{C_a p_{\text{escape}}(a; \Omega)}$$

where  $h$  is the unique function defined on  $\Omega \cup \partial\Omega$  that is harmonic in the set  $\Omega - \{a\}$  and satisfies the boundary conditions

$$(29) \quad h(a) = 1 \quad \text{and}$$

$$(30) \quad h(y) = 0 \quad \text{for all } y \in \partial\Omega.$$

*Proof.* Define a new graph  $\mathcal{G}_\Omega$  as follows:

- (1) Let the vertex set  $\mathcal{V}_\Omega = \Omega \cup \{b\}$ , where  $b$  is a new vertex.
- (2) Let the edge set  $\mathcal{E}_\Omega$  consist of
  - (a) all edges in the edge set  $\mathcal{E}$  with both endpoints in  $\Omega$ ; and
  - (b) one edge  $e_{\text{new}}$  with endpoints  $x$  and  $b$  for each edge  $e_{\text{old}}$  of  $\mathcal{E}$  with one endpoint  $x$  and the other in  $\partial\Omega$ .

Define a conductance function  $C^\Omega$  on this new graph  $\mathcal{G}_\Omega$  by setting  $C^\Omega(e) = C(e)$  for every edge  $e$  with both endpoints in  $\Omega$ , and setting  $C^\Omega(e_{\text{new}}) = C(e_{\text{old}})$  for every new edge  $e_{\text{new}}$  that corresponds to an edge  $e_{\text{old}}$  of  $\mathcal{G}$  connecting a vertex in  $\Omega$  to a vertex in  $\partial\Omega$ . Let  $p_{\text{escape}}^\Omega(a; b)$  be the escape probability for the Markov chain on  $\mathcal{G}_\Omega$  associated to the new conductance  $C^\Omega$ ; then clearly

$$p_{\text{escape}}(a; \Omega) = p_{\text{escape}}^\Omega(a; b).$$

But by Theorem 14 and Corollary 25,

$$p_{\text{escape}}^\Omega(a; b) = 1/(C_a R_{\text{eff}}^\Omega(a, b)) = 1/(C_a \min\{\mathcal{D}_\Omega(J, J) : J(a+) = J(b-) = 1\})$$

where the min is over all flows  $J$  in  $\mathcal{G}_\Omega$  with source  $a$ , sink  $b$ , and size 1.  $\square$

**5.4. Proof of Theorem 28.** Assume that there is a finite energy flow  $I$  on  $\mathcal{G}$  with source  $a$ . Without loss of generality we may assume that this is a *unit flow*, i.e., that  $I(a+) = 1$ . We will show that

$$P^a\{\text{return to } a\} < 1.$$

Let  $\Omega_1 \subseteq \Omega_2 \subseteq \dots$  be finite subsets of  $\mathcal{V}$ , each containing  $a$ , whose union is all of  $\mathcal{V}$ . Clearly the escape probabilities  $p_{\text{escape}}(a; \Omega_n)$  are monotone, and converge to the probability of no return to  $a$ . Thus, it suffices to prove that the escape probabilities are bounded away from 0.

By the Dirichlet Principle (the last corollary) the escape probabilities are given by

$$p_{\text{escape}}(a; \Omega_n) = \frac{1}{C_a \min \mathcal{D}_{\Omega_n}(J, J)},$$

where the min is over all unit flows on  $\Omega_n$  with source  $a$ . But this min is clearly no larger than the total energy of the flow  $I$ , because restricting  $I$  to  $\Omega_n$  gives a unit flow on  $\Omega_n$  with source  $a$ , and the energy of the restriction is no larger than the total energy of  $I$ . Therefore,

$$p_{\text{escape}}(a; \Omega_n) \geq \frac{1}{C_a \mathcal{D}(I, I)} > 0$$

for all  $n$ .  $\square$

**5.5. Proof of Theorem 29.** Assume that the Markov chain is transient. We will show that, for every vertex  $a \in \mathcal{V}$  such that the probability of no return to  $a$  is positive, there is a finite energy flow with source  $a$ . Let  $\Omega_1 \subseteq \Omega_2 \subseteq \dots$  be finite subsets of  $\mathcal{V}$ , each containing  $a$ , and whose union is all of  $\mathcal{V}$ . Define

$$\begin{aligned} h_n(x) &= P^x\{\text{hit } a \text{ before exiting } \Omega_n\}; \\ h(x) &= P^x\{\text{hit } a \text{ eventually}\}. \end{aligned}$$

Observe that  $h$  is harmonic at every  $x \in \mathcal{V} - \{a\}$ , and that  $h_n$  is harmonic at every  $x \in \Omega_n - \{a\}$ ; also, for every  $x$ ,  $h_n(x) \rightarrow h(x)$  as  $n \rightarrow \infty$ . Observe that the function  $h_n$  is the same function as appearing in the statement of Theorem 21.

Define flows  $I^{(n)}$  on  $\Omega_n$  and  $I$  on  $\mathcal{G}$  with source  $a$  by

$$\begin{aligned} I_{xy}^{(n)} &= C_{xy}(h_n(x) - h_n(y))/C_a p_{\text{escape}}(a; \Omega_n); \\ I_{xy} &= C_{xy}(h(x) - h(y))/C_a p_* \end{aligned}$$

where  $p_* = \downarrow \lim p_{\text{escape}}(a; \Omega_n) = P^a\{\text{no return to } a\} > 0$ . The flows  $I^{(n)}, I$  have source  $a$  and no other sources or sinks (by the harmonicity of  $h_n$  and  $h$  off  $a$ ). Moreover,  $I_{xy}^{(n)} \rightarrow I_{xy}$  as  $n \rightarrow \infty$ , for every directed edge  $xy$ .

By Proposition 24, the flows  $I^{(n)}$  minimize energy among all unit flows on  $\Omega_n$  with source  $a$ , and

$$p_* \leq p_{\text{escape}}(a; \Omega_n) = \frac{1}{C_a \mathcal{D}_{\Omega_n}(I^{(n)}, I^{(n)})}.$$

It follows that for every  $n$ ,

$$\mathcal{D}_{\Omega_n}(I^{(n)}, I^{(n)}) \leq \frac{1}{C_a p_*} < \infty.$$

But the energy of  $I$  is given by

$$\begin{aligned} \mathcal{D}(I, I) &= \sum_x \sum_y I_{xy}^2 R_{xy} \\ &= \sum_x \sum_y \lim_{n \rightarrow \infty} (I_{xy}^{(n)})^2 R_{xy} \\ &\leq \liminf_{n \rightarrow \infty} \sum_{x \in \Omega_n} \sum_y (I_{xy}^{(n)})^2 R_{xy} \\ &= \liminf_{n \rightarrow \infty} \mathcal{D}_{\Omega_n}(I^{(n)}, I^{(n)}) \\ &\leq \frac{1}{C_a p_*}, \end{aligned}$$

by the Fatou Lemma. Thus, the flow  $I$  has finite energy.  $\square$

## 6. THE TRANSFER CURRENT THEOREM

The *Transfer Current Theorem* of Burton and Pemantle relates the distribution of a random spanning tree to the electrical current flow(s) in the underlying graph. Since I can't follow the Burton-Pemantle proof, I am going to present my own.

Let  $\mathcal{G} = (V, \mathcal{E})$  be a finite, connected (multi-)graph (thus, multiple edges are allowed, but loops are not), and let  $\mathcal{E}^*$  be the set of *oriented* edges. Assume that each edge of  $\mathcal{G}$  is a resistor of conductance 1. For any two oriented edges  $\vec{e}_1, \vec{e}_2$ , let  $Y(\vec{e}_1, \vec{e}_2)$  be the (signed) electrical current that flows through  $\vec{e}_2$  when the terminals of a battery are attached to the tail  $x$  and head  $y$  of  $\vec{e}_1$  so that the resulting electrical current flow in the graph has source  $x$ , sink  $y$ , and size 1.

**Theorem 33.** *Let  $T$  be a random spanning tree, uniformly distributed on the set of all spanning trees of  $\mathcal{G}$ . For any edges  $e_1, e_2, \dots, e_k$ ,*

$$(31) \quad P\{e_1, e_2, \dots, e_k \in T\} = \det(Y(\vec{e}_i, \vec{e}_j))_{i,j=1,2,\dots,k},$$

where  $\vec{e}_i$  is any choice of orientations.

Observe that the special case  $k = 1$  has already been proved: it is a special case of Kirchhoff's theorem. This theorem states that for any two vertices  $x, y$  (not necessarily adjacent), when a

battery is attached to vertices  $x, y$  so that the resulting electrical current flow in the graph has source  $x$ , sink  $y$ , and size 1, the electrical current flow  $I^{xy}(\vec{e})$  in edge  $\vec{e}$  is given by

$$(32) \quad I^{xy}(\vec{e}) = \frac{N^{xy}(\vec{e}) - N^{xy}(-\vec{e})}{N},$$

Here  $N$  is the total number of spanning trees, and  $N^{xy}(\vec{e})$  is the number of spanning trees in which the unique path from  $x$  to  $y$  passes through  $e$  in the direction  $\vec{e}$ . In view of Kirchhoff's theorem, to prove Theorem 33 it suffices to prove that if relation (31) holds for  $k$ , then it holds for  $k+1$ .

**Lemma 34.** *For any two oriented edges  $\vec{e}, \vec{f}$ ,*

$$(33) \quad Y(\vec{e}, \vec{f}) = \frac{1}{2} \sum_{\vec{g} \in \mathcal{E}^*} Y(\vec{e}, \vec{g}) Y(\vec{f}, \vec{g}).$$

Consequently,

$$(34) \quad Y(\vec{e}, \vec{f}) = Y(\vec{f}, \vec{e}).$$

*Proof.* Relation (34) clearly follows from relation (33). To prove (33), one could rewrite  $Y(\vec{f}, \vec{g})$  as the gradient of a voltage function, using Ohm's Law, and then use the Conservation Law for flows. Here is another approach: By Ohm's Law,

$$\begin{aligned} Y(\vec{e}, \vec{g}) &= D^* \varphi_{\vec{e}}(\vec{g}) \quad \text{and} \\ Y(\vec{f}, \vec{g}) &= D^* \varphi_{\vec{f}}(\vec{g}), \end{aligned}$$

where  $D^*$  is the adjoint of the *incidence matrix*  $D$  of  $\mathcal{G}$ , defined by

$$\begin{aligned} D(x, \vec{e}) &= +1 & \text{if } x = \text{tail}(\vec{e}) \\ &= -1 & \text{if } x = \text{head}(\vec{e}) \\ &= 0 & \text{otherwise.} \end{aligned}$$

The incidence matrix is related to the combinatorial Laplacian  $\mathcal{L}$  by the equation

$$\mathcal{L} = \frac{1}{2} D D^*.$$

Consequently, if  $x$  and  $y$  are the tail and head of  $\vec{e}$ , respectively, then the inner product (33) may be rewritten as follows:

$$\begin{aligned} \frac{1}{2} \sum_{\vec{g} \in \mathcal{E}^*} Y(\vec{e}, \vec{g}) Y(\vec{f}, \vec{g}) &= (D^* \varphi_{\vec{e}}, D^* \varphi_{\vec{f}}) / 2 \\ &= (\varphi_{\vec{e}}, D D^* \varphi_{\vec{f}}) / 2 \\ &= (\varphi_{\vec{e}}, \mathcal{L} \varphi_{\vec{f}}) \\ &= (\varphi_{\vec{e}}, \delta_x - \delta_y) \\ &= \varphi_{\vec{e}}(x) - \varphi_{\vec{e}}(y) \\ &= D^* \varphi_{\vec{e}}(\vec{f}) \\ &= Y(\vec{e}, \vec{f}). \end{aligned}$$

□

**Lemma 35.** *Let  $e, f \in \mathcal{E}$  be two distinct edges of  $\mathcal{G}$ , and let  $\mathcal{H} = \mathcal{G}/f$  be the fused graph obtained from  $\mathcal{G}$  by deleting edge  $f$  and fusing the two endpoints of  $f$ . Let  $Y_{\mathcal{G}}$  and  $Y_{\mathcal{H}}$  denote the electrical current flows in the original graph  $\mathcal{G}$  and the fused graph  $\mathcal{H}$ , respectively. Then for any edge  $g \neq f$  and any choice of orientations,*

$$(35) \quad Y_{\mathcal{H}}(\vec{e}, \vec{g}) = Y_{\mathcal{G}}(\vec{e}, \vec{g}) - \frac{Y_{\mathcal{G}}(\vec{e}, \vec{f})}{Y_{\mathcal{G}}(\vec{f}, \vec{f})} Y_{\mathcal{G}}(\vec{f}, \vec{g}).$$

*Proof.* There are two ways to do this: (a) Use Ohm's Law to convert (35) to an equivalent statement about voltages, then use linearity and the uniqueness principle to prove it; or (b) Prove it directly. I'll follow strategy (b). Let  $x, y$  be the tail and head of  $\vec{e}$ , and let  $u, v$  be the tail and head of  $\vec{f}$ . To prove the equality (35), it suffices to show that the right side defines an electrical current flow in  $\mathcal{H}$  with source  $x$ , sink  $y$ , and size 1. For this it suffices to prove (1) that the right side satisfies the flow equation at every vertex  $z$  of  $\mathcal{H}$  other than  $x, y$ ; and (2) that the right side defines an anti-symmetric *acyclic* function on the directed edges of  $\mathcal{H}$ .

(1) The function  $Y_{\mathcal{G}}(\vec{e}, \cdot)$  is a flow with source  $x$  and sink  $y$ , and the function  $Y_{\mathcal{G}}(\vec{f}, \cdot)$  is a flow with source  $u$  and sink  $v$ . The right side of (35) is a linear combination of these two functions. As such, it is antisymmetric on the set of directed edges, and it is a flow with *possible* sinks and sources  $x, u$  and  $y, v$ , respectively. Note, however, that the multiplier has been chosen so that

$$Y_{\mathcal{G}}(\vec{e}, \vec{f}) - \frac{Y_{\mathcal{G}}(\vec{e}, \vec{f})}{Y_{\mathcal{G}}(\vec{f}, \vec{f})} Y_{\mathcal{G}}(\vec{f}, \vec{f}) = 0,$$

so that there is *zero* flow across the edge  $f$ . Thus, the flow projects to a flow on the fused graph  $\mathcal{H}$ .

(2) Any cycle in  $\mathcal{H}$  is the projection of a cycle in  $\mathcal{G}$ . Since any flow on  $\mathcal{G}$  is acyclic, and since the right side of (35) is a linear combination of two flows on  $\mathcal{G}$ , it follows that the projection to  $\mathcal{H}$  is also acyclic.  $\square$

**Lemma 36.** *Let  $A = (a_{ij})$  be a symmetric  $(n+1) \times (n+1)$  matrix indexed by  $i, j = 0, 1, \dots, n$ . Let  $B = (b_{ij})$  be the symmetric  $n \times n$  matrix indexed by  $i, j = 1, 2, \dots, n$  whose entries are*

$$(36) \quad b_{ij} = a_{ij} - a_{i0}a_{0j}/a_{00}.$$

*Then*

$$(37) \quad \det A = a_{00} \det B$$

*Proof.* This is (probably) a standard theorem in matrix theory. It follows by expanding  $\det A$  by minors (across the top row).  $\square$

*Proof.* Proof of Theorem 33 We will use induction on  $k$ . By Kirchhoff's Theorem, the formula is true when  $k = 1$ . Assume that it is true for  $k \geq 1$ ; we will prove that it must then be true for  $k+1$ . Let  $e_0, e_1, e_2, \dots, e_k$  be  $k+1$  edges of the graph  $\mathcal{G}$ . Let  $\mathcal{H} = \mathcal{G}/e_0$  be the fused graph obtained by deleting edge  $e_0$  and fusing its endpoints. Let  $T$  be a random spanning tree of  $\mathcal{G}$ , and let  $T'$  be a random spanning tree of  $\mathcal{H}$ . Then

$$\begin{aligned} P(e_0, e_1, e_2, \dots, e_k \in T) &= P(e_0 \in T)P(e_1, e_2, \dots, e_k \in T \mid e_0 \in T) \\ &= P(e_0 \in T)P(e_1, e_2, \dots, e_k \in T'). \end{aligned}$$

But by the induction hypothesis, for any orientation of the edges  $e_i$ ,

$$P(e_0 \in T) = Y_g(\vec{e}_0, \vec{e}_0)$$

and

$$P(e_1, e_2, \dots, e_k \in T') = \det(Y_{\mathcal{H}}(\vec{e}_i, \vec{e}_j))_{i,j=1,2,\dots,k}.$$

By Lemma 35,

$$Y_{\mathcal{H}}(\vec{e}_i, \vec{e}_j) = Y_g(\vec{e}_i, \vec{e}_j) - \frac{Y_g(\vec{e}_i, \vec{e}_0)}{Y_g(\vec{e}_0, \vec{e}_0)} Y_g(\vec{e}_0, \vec{e}_j).$$

Consequently, the formula (31) follows from Lemma 36. □