# Unimodularity and Stochastic Processes 

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We might explain unimodularity as a non-obvious use of group-invariance. Simplest setting: transitive graphs. A graph is a pair $G=(\mathrm{V}, \mathrm{E})$ with E a symmetric subset of $\mathrm{V} \times \mathrm{V}$. An automorphism of $G$ is a permutation of V that induces a permutation of E. The set of all automorphisms of $G$ forms a group, $\operatorname{Aut}(G)$. We call $G$ transitive if Aut $(G)$ acts transitively on $\vee$ (i.e., there is only one orbit).


(Don Hatch)

Consider the following examples: Let $G$ be an infinite transitive graph and let $\mathbf{P}$ be an invariant percolation, i.e., an $\operatorname{Aut}(G)$-invariant measure on $2^{\vee}$, on $2^{\mathrm{E}}$, or even on $2^{\mathrm{V} \cup E}$. Let $\omega$ be a configuration with distribution $\mathbf{P}$.

Example: Could it be that $\omega$ is a single vertex? I.e., is there an invariant way to pick a vertex at random?

No: If there were, the assumptions would imply that the probability $p$ that $\omega=\{x\}$ is the same for all $x$, whence an infinite sum of $p$ would equal 1 , an impossibility.

Example: Could it be that $\omega$ is a finite nonempty vertex set? I.e., is there an invariant way to pick a finite set of vertices at random?

No: If there were, then we could pick one of the vertices of the finite set at random (uniformly), thereby obtaining an invariant probability measure on singletons.

Cluster means connected component of $\omega$. A vertex $x$ is a furcation of a configuration $\omega$ if removing $x$ would split the cluster containing $x$ into at least 3 infinite clusters.


Example: Is the number of furcations $\mathbf{P}$-a.s. 0 or $\infty$ ? Yes, for the set of furcations has an invariant distribution on $2^{\vee}$.

Example: Does P-a.s. each cluster have 0 or $\infty$ furcations?
This does not follow from elementary considerations as the previous examples do (we can prove this).

But suppose we have the following kind of conservation of mass.
We call $f: \vee \times \vee \rightarrow[0, \infty]$ diagonally invariant if $f(\gamma x, \gamma y)=f(x, y)$ for all $x, y \in \mathrm{~V}$ and $\gamma \in \operatorname{Aut}(G)$.

The Mass-Transport Principle. For all diagonally invariant $f$, we have

$$
\sum_{x \in \mathrm{~V}} f(o, x)=\sum_{x \in \mathrm{~V}} f(x, o),
$$

where $o$ is any fixed vertex of $G$.

Suppose this holds.

Write $K(x)$ for the cluster containing $x$.
Now, given the configuration $\omega$, define $F(x, y ; \omega)$ to be 0 if $K(x)$ has 0 or $\infty$ furcations, but to be $1 / N$ if $y$ is one of $N$ furcations of $K(x)$ and $1 \leq N<\infty$. Then $F$ is diagonally invariant, whence the Mass-Transport Principle applies to $f(x, y):=\mathbf{E} F(x, y ; \omega)$. Since $\sum_{y} F(x, y ; \omega) \leq 1$, we have

$$
\begin{equation*}
\sum_{x} f(o, x) \leq 1 \tag{1}
\end{equation*}
$$

If any cluster has a finite positive number of furcations, then each of them receives infinite mass. More precisely, if $o$ is one of a finite number of furcations of $K(o)$, then $\sum_{x} F(x, o ; \omega)=\infty$. Therefore, if with positive probability some cluster has a finite positive number of furcations, then with positive probability $o$ is one of a finite number of furcations of $K(o)$, and therefore $\mathbf{E}\left[\sum_{x} F(x, o ; \omega)\right]=\infty$. That is, $\sum_{x} f(x, o)=\infty$, which contradicts the Mass-Transport Principle and (1).

Call $G$ unimodular if the Mass-Transport Principle holds for $G$. Which graphs enjoy this wonderful property? All graphs do that are properly embedded in euclidean or hyperbolic space with a transitive action of isometries of the space. All Cayley graphs do:

We say that a group $\Gamma$ is generated by a subset $S$ of its elements if the smallest subgroup containing $S$ is all of $\Gamma$. In other words, every element of $\Gamma$ can be written as a product of elements of the form $s$ or $s^{-1}$ with $s \in S$. If $\Gamma$ is generated by $S$, then we form the associated Cayley graph $G$ with vertices $\Gamma$ and (unoriented) edges $\left\{(x, x s) ; x \in G, s \in S \cup S^{-1}\right\}$. Because $S$ generates $\Gamma$, the graph is connected. Cayley graphs are transitive since left multiplication by $y x^{-1}$ is an automorphism of $G$ that carries $x$ to $y$.

Now if $f: \Gamma^{2} \rightarrow[0, \infty]$ is diagonally invariant, then for $o$ the identity of $\Gamma$ and any $x \in \Gamma$, we have $f(o, x)=f\left(x^{-1}, o\right)$. Since inversion preserves counting measure on $\Gamma$, we obtain the Mass-Transport Principle.
(For a general transitive graph, the Mass-Transport Principle is equivalent to unimodularity of Haar measure on $\operatorname{Aut}(G)$. History: Liggett (1985), Adams (1990), van den Berg and Meester (1991), Häggström (1997), Benjamini, L., Peres, Schramm (1999). I ignore other uses of unimodularity in probability that go back considerably longer.)

Non-example: the "grandparent" graph of Trofimov:


The grandparent graph is not unimodular: let $f(x, y)$ be the indicator that $y$ is the grandparent of $x$. Then

$$
\sum_{x} f(o, x)=1
$$

while

$$
\sum_{x} f(x, o)=4
$$

Another definition: $G$ is amenable if there is a sequence $K_{n}$ of finite vertex sets in $G$ such that the number of neighbors of $K_{n}$ divided by the size of $K_{n}$ tends to 0 .

Example: $\mathbb{Z}^{d}$
Non-examples: regular trees of degree at least 3; hyperbolic tessellations.
All amenable transitive graphs are unimodular (Soardi and Woess).

A selection of theorems:
$\operatorname{Bernoulli}(p)$ percolation on $G$ puts each edge in $\omega$ independently with probability $p$. The probability of an infinite cluster in $\omega$ is 0 or 1 by Kolmogorov's $0-1$ Law. It increases in $p$, so there is a critical value $p_{\mathrm{c}}$ where it changes. What is the probability of an infinite cluster at $p_{\mathrm{c}}$ ? Benjamini and Schramm conjectured it is 0 on any transitive graph, provided that $p_{\mathrm{c}}<1$. It was known for $\mathbb{Z}^{d}$ for $d=2$ (Kesten) and $d \geq 19$ (Hara and Slade).

THEOREM (BLPS 1999). This is true for all non-amenable transitive unimodular graphs.

It is unknown whether this holds for non-unimodular graphs.

Theorem (Häggström; HÄggström and Peres; L. and Peres; L. and Schramm). Let $G$ be a transitive unimodular graph. Given invariant random transition probabilities $p_{\omega}(x, y)$ and an invariant p-stationary measure $\nu_{\omega}(x)$, biasing $\omega$ by $\nu_{\omega}(o)$ gives a measure that is invariant from the point of view of the walker.

Example: Degree-biasing for simple random walk on the clusters.
This is false on non-unimodular graphs.


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Theorem (Aldous and L.). Let $G$ be a transitive unimodular graph. Given invariant random symmetric rates $r_{\omega}(x, y)$ such that $\mathbf{E}\left[\sum_{x} r(o, x)\right]<\infty$, the associated continuous-time random walk has no explosions a.s.

This is false on non-unimodular graphs.


Proof. Let $Z:=\mathbf{E}\left[\sum_{x} r(o, x)\right]$. Consider the discrete-time random walk $\left\langle X_{n} ; n \geq 0\right\rangle$ corresponding to the weights $r_{\omega}(x, y)$. This has a stationary measure

$$
\nu_{\omega}(x):=\sum_{y} r_{\omega}(x, y) / Z
$$

It also describes the steps of the continuous-time random walk, ignoring the waiting times. Biasing by $\nu_{\omega}$ gives a probability measure. The continuous-time random walk moves from $x$ at rate $\sum_{y} r_{\omega}(x, y)=Z \nu_{\omega}(x)$, so spends expected time $1 /\left(Z \nu_{\omega}(x)\right)$ before moving (given $\omega$ ). Thus, it explodes w.p.p. given $\left\langle X_{n}\right\rangle$ iff $\sum_{n} 1 / \nu_{\omega}\left(X_{n}\right)<\infty$ by the Borel-Cantelli lemma. But this sum is infinite by stationarity and Poincaré's recurrence theorem (which applies because the biased measure is finite).

Theorem (Fontes and Mathieu; Aldous and L.). Let $G$ be a transitive unimodular graph. Given invariant random pairs of symmetric rates $\left(r_{\omega}, R_{\omega}\right)$ such that

$$
r_{\omega}(x, y) \leq R_{\omega}(x, y)
$$

for all $x, y$ and almost all $\omega$, let $p_{t}(o, o)$ and $P_{t}(o, o)$ be the expected [annealed] return probabilities for the associated continuous-time (minimal) random walks. Then for all $t>0$, we have

$$
p_{t}(o, o) \geq P_{t}(o, o) .
$$

It is unknown whether this holds for non-unimodular graphs. It is also unknown if we assume invariance of $r_{\omega}$ and $R_{\omega}$ separately.

Proof. Let

$$
a_{\omega}(x, y):= \begin{cases}-r_{\omega}(x, y) & \text { if } x \neq y \\ \sum_{z} r_{\omega}(x, z) & \text { if } x=y\end{cases}
$$

and

$$
A_{\omega}(x, y):= \begin{cases}-R_{\omega}(x, y) & \text { if } x \neq y \\ \sum_{z} R_{\omega}(x, z) & \text { if } x=y\end{cases}
$$

Then

$$
p_{t}^{\omega}(x, y)=\left(e^{-t a_{\omega}}\right)(x, y) \text { and } P_{t}^{\omega}(x, y)=\left(e^{-t A_{\omega}}\right)(x, y)
$$

Thus,

$$
p_{t}(o, o)=\mathbf{E}\left[e^{-t a_{\omega}}(o, o)\right]=: \operatorname{Tr}\left[e^{-t a_{\omega}}\right]
$$

Since $\left[\begin{array}{cc}1 & -1 \\ -1 & 1\end{array}\right] \geq 0$, we have $a_{\omega} \leq A_{\omega}$. Therefore

$$
p_{t}(o, o)=\operatorname{Tr}\left[e^{-t a_{\omega}}\right] \geq \operatorname{Tr}\left[e^{-t A_{\omega}}\right]=P_{t}(o, o)
$$

Extensions of unimodularity:
On finite graphs, the Mass-Transport Principle is obvious if we take $o$ to be a uniform random "root" and average over $o$ :

$$
\begin{equation*}
\mathbf{E}\left[\sum_{x} f(o, x)\right]=\mathbf{E}\left[\sum_{x} f(x, o)\right] . \tag{2}
\end{equation*}
$$

This is just interchanging the order of summation. But it is crucial that the root be chosen uniformly. Indeed, (2) characterizes the uniform measure.

Consider this graph:


We should choose $o$ to be a blue vertex with probability twice that of a black vertex in order that (2) hold.

With this graph:

we should choose $o$ to be a blue vertex with probability four times that of a black vertex in order that the Mass-Transport Principle hold,

$$
\mathbf{E}\left[\sum_{x} f(o, x)\right]=\mathbf{E}\left[\sum_{x} f(x, o)\right] .
$$

What about the hyperbolic triangle tessellation?


We call $G$ quasi-transitive if $\operatorname{Aut}(G)$ acts quasi-transitively on V (i.e., there are only finitely many orbits). If $G$ is quasi-transitive and amenable, then each orbit has a natural frequency (BLPS), which should be used for the probability of choosing a representative from that orbit for $o$ in the Mass-Transport Principle,

$$
\mathbf{E}\left[\sum_{x} f(o, x)\right]=\mathbf{E}\left[\sum_{x} f(x, o)\right] .
$$

If there are probabilities $\alpha_{i}$ for the orbit representatives $o_{1}, \ldots, o_{L}$ such that choosing $o_{i}$ with probability $\alpha_{i}$ makes the Mass-Transport Principle true, then we call $G$ unimodular.

How do we tell? The following is necessary and sufficient: if $x$ is in the orbit of $o_{i}$ and $y$ is in the orbit of $o_{j}$, then

$$
\frac{|S(x) y|}{|S(y) x|}=\frac{\alpha_{j}}{\alpha_{i}}
$$

where $S(x):=\{\gamma \in \operatorname{Aut}(G) ; \gamma x=x\}$.

Consider now the space of rooted graphs or networks. In fact, consider only rootedisomorphism classes of networks. A probability measure on this space is unimodular if the Mass-Transport Principle holds:

$$
\begin{equation*}
\mathbf{E}\left[\sum_{x \in \mathrm{~V}(G)} f(G ; o, x)\right]=\mathbf{E}\left[\sum_{x \in \mathrm{~V}(G)} f(G ; x, o)\right] \tag{3}
\end{equation*}
$$

for all Borel non-negative $f$ that are diagonally invariant under isomorphisms.

For example, as observed by Benjamini/Schramm and by Aldous/Steele, all weak limits of uniformly rooted finite networks are unimodular.

All the theorems given for transitive unimodular graphs hold for unimodular random rooted networks (Aldous-L.).

Example: If we want the offspring distribution $\left\langle p_{k}\right\rangle$ for a unimodular version UGW of Galton-Watson trees, let $r_{k}:=c^{-1} p_{k-1} / k$ for $k \geq 1$ and $r_{0}:=0$, where $c:=$ $\sum_{k \geq 0} p_{k} /(k+1)$. With the sequence $\left\langle r_{k}\right\rangle$ and $n$ vertices, give each vertex $k$ balls with probability $r_{k}$, independently. Then pair the balls at random and place an edge for each pair between the corresponding vertices. There may be one ball left over; if so, ignore it. In the limit, we get a random tree where the root has degree $k$ with probability $r_{k}$ and each neighbor of the root has an independent Galton-Watson $\left(\left\langle p_{k}\right\rangle\right)$ tree.

$\left(150\right.$ vertices with $\left.p_{1}=p_{2}=1 / 2\right)$


Example: Biasing UGW by the degree of the root gives a stationary measure for simple random walk (L., Pemantle and Peres):


Example: Aperiodic tessellations. Like Palm measure.


