

The Frog Problem

Huy Mai

Brandeis University

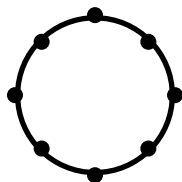
Gordon Rojas Kirby

Stanford University

PSU Math REU

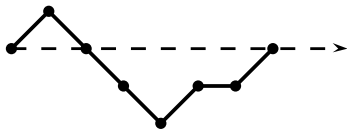
August 7, 2012

Preliminaries

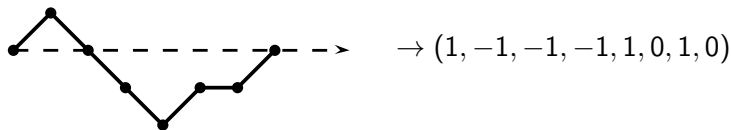


- ▶ Cycle graph, such that each frog has two neighbors
- ▶ Identical probability p for each frog at each iteration
- ▶ Restriction: A frog returns if it is more than one level above its neighbors

Preliminaries



Preliminaries



- ▶ Represent Motzkin Path as a vector
- ▶ We call this a **state**
- ▶ The number of states are enumerated by the central trinomial coefficients
- ▶ **Our goal is to get a bound for the asymptotic speed $S(K)$**

Markov Chains

Definition

A **Markov Chain** is collection of random variables X_t , $t = 0, 1, \dots$, having the property that, given the present, the future is conditionally independent of the past.

Markov Chains

Definition

A **Markov Chain** is collection of random variables X_t , $t = 0, 1, \dots$, having the property that, given the present, the future is conditionally independent of the past.

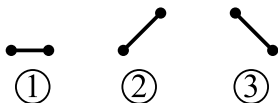
- ▶ For any finite K we have a Markov chain where the random variable represent transition probabilities between states

Example, $K = 2$



$$M = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{pmatrix} p_{11} & p_{12} & p_{13} \\ p_{21} & p_{22} & p_{23} \\ p_{31} & p_{32} & p_{33} \end{pmatrix} \end{matrix}$$

Example, $K = 2$



$$M = \begin{pmatrix} Q & \vdots & R \\ \dots & & \dots \\ 0 & \vdots & I_K \end{pmatrix}$$

For $K = 2$ we have

$$Q = \begin{pmatrix} q^2 & pq & pq \\ 0 & pq + q^2 & 0 \\ 0 & 0 & pq + q^2 \end{pmatrix} \quad \text{and} \quad R = \begin{pmatrix} p^2 & 0 & 0 \\ pq & p^2 & 0 \\ pq & 0 & p^2 \end{pmatrix}.$$

Example, $K = 2$

It can be easily checked that

$$(I - Q)^{-1} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} \frac{2p-3}{p(p-2)} \\ \frac{1}{2p(1-p)} \\ \frac{1}{2p(1-p)} \end{pmatrix}$$

Example, $K = 2$

It can be easily checked that

$$(I - Q)^{-1} \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} = \begin{pmatrix} \frac{2p-3}{p(p-2)} \\ \frac{1}{2p(1-p)} \\ \frac{1}{2p(1-p)} \end{pmatrix}$$

- ▶ We can do this for any fixed $K \in \mathbb{N}$, i.e. $S(K)$ exists
- ▶ In this example we see that the expected time spent in state 2 and state 3 is the same

- ▶ e.g. When $p = \frac{1}{2}$ we have $\begin{pmatrix} 8/3 \\ 2 \\ 2 \end{pmatrix}$

We need a new approach for when $K \rightarrow \infty$. Define

- ▶ $X_{n,k}$ are our random variables indexed by level and frog
- ▶ Indicator function $L_n(i, j)$
- ▶ $T(N, K) = \max_{k_1, \dots, k_N} \{ (\prod_{n=2}^N L_n(k_{n-1}, k_n)) \sum_{n=1}^N X_{n, k_n} \}$
- ▶ $r = E[\sum_{j=1}^K L_n(i, j)]$ for all i and n

Theorem (Chang and Nelson)

If the moment generating function of the time it takes for a frog to jump, $X_{n,k}$, exists for some finite $0 < \theta_0$

$$\phi(\theta) \stackrel{\text{def}}{=} E[e^{\theta X_{n,k}}] < \infty \quad \text{for } \theta \leq \theta_0$$

then the asymptotic speed $S(K)$ for all of the frogs to jump higher than some fixed level is bounded below $\frac{1}{t^*}$, where

$$t^* = \inf\{t \geq 1 \mid rm(t) < 1\}$$

and

$$m(t) = \inf_{0 < \theta < \theta_0} \{e^{-\theta t} \phi(\theta)\}.$$

Sketch of Proof

Definition

A **Martingale** is a sequence of random variables X_1, X_2, \dots where the following is true

$$E[|X_n|] < \infty \quad \text{and} \quad E[X_{n+1}|X_1, \dots, X_n] = X_n$$

$$M_n(\theta) = \frac{1}{(r\phi(\theta))^n} \sum_{k_1=1}^K \dots \sum_{k_N=1}^K \left(\prod_{m=2}^n L_{m-1}(k_{m-1}, k_m) \right) e^{\theta \sum_{m=1}^n X_{m,k_m}}$$

Sketch of Proof

Lemma

For the system as previously defined,

$$\frac{1}{S(K)} = \limsup_{N \rightarrow \infty} \frac{T(N, K)}{N} \leq t^*, \quad \text{a.s.}$$

$$e^{\theta T(N, K)} = \max_{k_1, \dots, k_N} \left\{ \prod_{n=2}^N L_n(k_{n-1}, k_n) e^{\theta \sum_{n=1}^N X_{n, k_n}} \right\} \leq (r\phi(\theta))^N M_N(\theta).$$

Sketch of Proof

Markov's Inequality

$$P(X > t) \leq \frac{E[X]}{t}$$

$$\begin{aligned} P\left(\frac{T(N, K)}{N} > t\right) &= P\left(e^{\theta T(N, K)} > e^{\theta Nt}\right) \\ &\leq e^{-\theta Nt} E\left[e^{\theta T(N, K)}\right] \\ &\leq \frac{K}{r} (re^{-\theta t} \phi(\theta))^N \end{aligned}$$

Sketch of Proof

Thus, since $\sum_N P\left(\frac{T(N,K)}{N} > t\right) < \infty$ if $rm(t) < 1$. Hence we have

$$P\left(\limsup_{N \rightarrow \infty} \left(\frac{T(N,K)}{N} > t\right)\right) = 0$$

which implies

$$\limsup_{n \rightarrow \infty} \frac{T(N,K)}{N} \leq t^*, \text{ a.s}$$

Main Theorem

Theorem

Given K^2 iterations, the probability that all frogs have jumped above level N , where $N = aK^2$ and $a < \frac{-\ln q}{\ln r}$, is $1 - \alpha\beta^{K^2}$, where $\beta < 1$. That is,

$$P(T(N, K) \leq K^2) \xrightarrow{K \rightarrow \infty} 1$$

Main Theorem

Theorem

Given K^2 iterations, the probability that all frogs have jumped above level N , where $N = aK^2$ and $a < \frac{-\ln q}{\ln r}$, is $1 - \alpha\beta^{K^2}$, where $\beta < 1$. That is,

$$P(T(N, K) \leq K^2) \xrightarrow{K \rightarrow \infty} 1$$

$$P(T(N, K) > K^2) = P(e^{\theta T(N, K)} > e^{\theta K^2})$$

$$\phi(\theta) = \frac{pe^{\theta}}{1 - qe^{\theta}}$$

Acknowledgements

We would like to thank Misha Guysinski and the support and funding we received from the PSU Math REU

Huy Mai

Brandeis University

huymai@brandeis.edu

Gordon Rojas Kirby

Stanford University

girkirby@stanford.edu