

can require a number of spin flips that is exponential in N (A. Haken et al. ca. 1989), and that one can in fact embed arbitrary computations in the dynamics (Orponen 1995). (More precisely, determining whether a given “output spin” is $+1$ or -1 in the local minimum reached from a given initial state is a “PSPACE-complete” problem.)

6.4 The NK Model

Introduced by Stuart Kauffman (ca. 1986) as a “tunable family of fitness landscapes”.

A *fitness landscape* is a triple $\langle X, R, f \rangle$, where X is the *configuration* (or *state*) *space*, $R \subseteq X \times X$ is a *neighbourhood relation* on X , and $f : X \rightarrow \mathbb{R}$ is a *fitness* (or *objective*) *function*.

A point $x \in X$ is a *local optimum* (of f on X) if

$$f(y) \leq f(x) \quad \forall yRx$$

and a *global optimum* (*maximum*) if

$$f(y) \leq f(x) \quad \forall y \in X$$

Questions of the “ruggedness” of landscapes (correlation structure), number and height of local optima, sizes of “attraction basins” of local optima with respect to “hill-climbing” algorithms etc. are of great interest for natural landscapes.

In Kauffman’s NK models, $X = A^N$ (usually just $X = \{0, 1\}^N$) and K is a tunable neighbourhood size parameter that influences the landscape characteristics, especially its ruggedness (cf. Figure 4).

The model can be seen as a toy model of “epigenetic interactions in chromosomes” — or also a generalisation of the spin glass model.

In Kauffman’s model, a *chromosome* is a N -vector of *loci* (*genes*, “positions”), each of which has a value from a set of *alleles* A (usually just $A = \{0, 1\}$). A “filled-in” chromosome $\alpha \in A^N$ is called a *genotype*.

The fitness of each gene $i \in \{1, \dots, N\}$ in a genotype $\alpha = (a_1, \dots, a_N) \in A^N$ depends on the allele a_i and K other alleles a_1^i, \dots, a_K^i via some local fitness function $f^i(\alpha) = f^i(a_i; a_1^i, \dots, a_K^i)$, usually normalised so that $f^i(\alpha) \in [0, 1]$. The total fitness of a genotype $\alpha \in A^N$ is the normalised sum of its genes’ local fitnesses:

$$f(\alpha) = \frac{1}{N} \sum_{i=1}^N f^i(a_i; a_1^i, \dots, a_K^i) \in [0, 1].$$

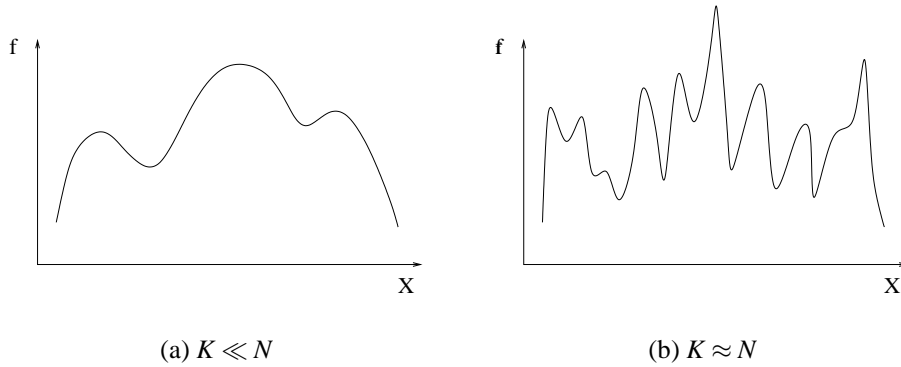


Figure 4: A smooth (a) and a rugged (b) NK fitness landscape.

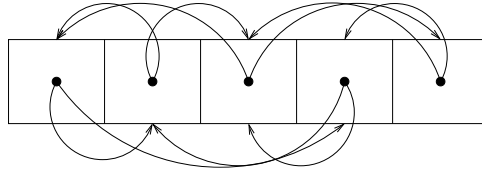


Figure 5: An NK interaction network with $N = 5$, $K = 2$.

Figure 5 illustrates an NK network with five loci and two “epigenetic interactions” per locus.

In Kauffman’s versions of the model, the K loci affecting locus i can either be systematically selected as e.g. $i + 1, \dots, i + K \pmod{N}$, or the chromosome can be simply “randomly wired”. The f^i are usually determined as randomly generated 2^{K+1} -element “interaction tables”.

From the spin glass perspective, e.g. a 1-D Ising model with N spins can be seen as an N^2 network where $f^i(S_i; S_{i-1}, S_{i+1}) = \frac{J}{2}(S_{i-1}S_i + S_iS_{i+1})$, and an SK spin glass with coefficients J_{ij} and local fields h_i as an $N(N - 1)$ network where

$$f^i(S_i; \sigma \setminus \{S_i\}) = \frac{1}{2} \sum_{\langle ij \rangle} J_{ij} S_i S_j + h_i S_i.$$

Basic properties of the NK model, for binary alleles $A = \{0, 1\}$ and varying values of K , include the following:

$K = 0$:

If $f^i(0) \neq f^i(1) \forall i = 1, \dots, N$, then there is a unique global optimum, which is easily found by e.g. the obvious 1-locus mutation “hill-climbing” algorithm.

Expected length of the hill-climbing path is $N/2$. (Half of the alleles are “right” in the beginning, after that one allele gets fixed at each step.)

Neighbouring genotypes α, α' are always highly correlated, as necessarily $|f(\alpha) - f(\alpha')| \leq 1/N$.

$1 \leq K < N - 1$:

For $K = 1$, a global optimum can still be found in polynomial time. For $K \geq 2$, global optimisation is NP-complete. However, for adjacent affecting loci ($i \curvearrowright i + 1, \dots, i + K$), the problem can be solved in time $o(2^K N)$ (Weinberger).

$K = N - 1$:

Neighbouring genotypes are totally uncorrelated.

\Rightarrow Probability that a given genotype α is a local optimum is equal to the probability that α has the highest rank within its 1-mutant neighbourhood. This probability is equal to $1/(N + 1)$.

\Rightarrow The expected number of local optima is $2^N/(N + 1)$.

The expected number of improvement steps for 1-mutant hill-climbing to hit a local optimum is proportional to $\log_2 N$ (each improvement step typically halves the rank of the genotype within the neighbourhood).

The expected waiting time for finding an improvement step is proportional to N .

7 Random Graphs

7.1 The Erdős-Rényi Model(s)

Two closely related “uniform” random graph models introduced in 1959 by P. Erdős & A. Rényi and E. N. Gilbert.

Consider the family \mathcal{G}_n of all (labelled, undirected) graphs on n nodes. Denote $N = \binom{n}{2}$; then $|\mathcal{G}_n| = 2^N$.

Define the following two probability spaces

[Erdős & Rényi:] $\mathcal{G}(n, M) =$ all $G \in \mathcal{G}_n$ with exactly $M \leq N$ edges, taken with uniform probability, i.e.

$$\Pr(G_M = H) = \begin{cases} \binom{N}{M}^{-1}, & \text{if } H \text{ has } M \text{ edges} \\ 0; & \text{otherwise.} \end{cases}$$

[Gilbert:] $\mathcal{G}(n, p) =$ all $G \in \mathcal{G}_n$, taken so that each edge has occurrence probability p , $0 \leq p \leq 1$, independently of the other edges, i.e.

$$\Pr(G_p = H) = p^M \underbrace{(1-p)^{N-M}}_q, \text{ if } H \text{ has } M \text{ edges.}$$

These spaces are in a precise sense “close” if $M \sim pN$, and are often both referred to (unfairly to Gilbert) as the “Erdős-Rényi random graph model”, or alternatively as the $\mathcal{G}(n, M)$ and $\mathcal{G}(n, p)$ random graph models.

Let $\Omega_n, n = 0, 1, 2, \dots$ be a sequence of probability spaces of n -node graphs. Say that *almost every* (a.e.) graph in Ω_n has property Q if

$$\Pr(G \in \Omega_n \text{ has } Q) \rightarrow 1, \text{ as } n \rightarrow \infty.$$

Conversely, *almost no* graph in Ω_n has property Q if a.e. graph in Ω_n has property $\neg Q$, i.e.

$$\Pr(G \in \Omega_n \text{ has } Q) \rightarrow 0, \text{ as } n \rightarrow \infty.$$

Theorem 7.1 *Let H be a fixed graph and p a constant, $0 < p < 1$. Then a.e. $G \in \mathcal{G}(n, p)$ contains an induced copy of H .*

Remark: an “induced copy” means here a subset of nodes whose induced subgraphs is isomorphic to H .

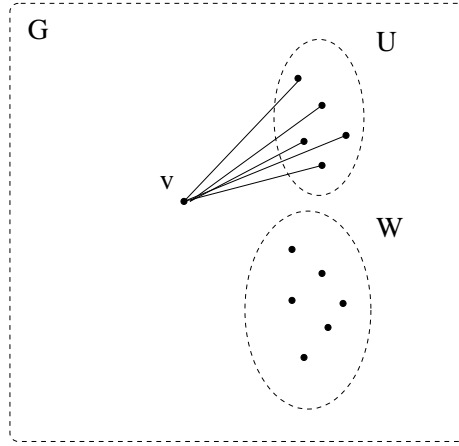
Proof. Let $k = |H| =$ number of nodes in H . Then a graph G with $n = |G| \geq k$ nodes can be partitioned into $\lfloor n/k \rfloor$ disjoint sets of k nodes (with some left over). For each of these sets, the probability that it forms an induced copy of H is $r > 0$. (Precisely, $r = \frac{k!}{|\text{Aut}(H)|} p^{e(H)} q^{\binom{k}{2} - e(H)}$.)

Thus, the probability that none of these sets forms an induced copy of H is

$$(1-r)^{\lfloor n/k \rfloor} \rightarrow 0, \text{ as } n \rightarrow \infty. \square$$

Let $k, l \in \mathbb{N}$. Say that a graph $G = (V, E)$ has property Q_{kl} if $\forall U, W, |U| \leq k, |W| \leq l, U \cap W = \emptyset, G$ contains a node $v \in V$ such that v is adjacent to all $u \in U$ and no $w \in W$ (cf. Figure 6).

Lemma 7.2 *For every constant p , $0 < p < 1$, and all $k, l \in \mathbb{N}$, a.e. $G \in \mathcal{G}(n, p)$ has property Q_{kl} .*

Figure 6: Property Q_{kl} .

Proof. For a fixed $U, W, v \in V \setminus (U \cup W)$, the probability that the condition is satisfied is

$$p^{|U|} q^{|W|} \geq p^k q^l$$

The events are independent for different v , so the probability that no appropriate v exists is

$$\left(1 - p^{|U|} q^{|W|}\right)^{n - |U| - |W|} \leq \left(1 - p^k q^l\right)^{n - k - l}.$$

There are at most n^{k+l} (U, W) -pairs to be considered, so the probability that some pair has no good v is bounded by

$$n^{k+l} \underbrace{\left(1 - p^k q^l\right)^{n - k - l}}_{< 1} \rightarrow 0, \text{ as } n \rightarrow \infty.$$

Thus in a.e. $G \in \mathcal{G}(n, p)$ all (U, W) -pairs have some appropriate v . \square

Corollary 7.3 *Let p , $0 < p < 1$, be a constant. Then (i) a.e. $G \in \mathcal{G}(n, p)$ has minimum degree $\geq k$, for given constant k (ii) a.e. $G \in \mathcal{G}(n, p)$ has diameter 2 (iii) a.e. $G \in \mathcal{G}(n, p)$ is k -connected for given constant k .*

Proof. (i) and (ii) are immediate.

(iii) In a.e. $G \in \mathcal{G}(n, p)$, no two nodes u_1, u_2 can be separated by a cutset of size $k - 1$, because we may choose in Lemma 7.2 $U = u_1, u_2$, $W = w_1, \dots, w_{k-1}$ for arbitrary w_1, \dots, w_{k-1} , and obtain a path $u_1 - v - u_2$ connecting u_1, u_2 and avoiding w_1, \dots, w_{k-1} . \square

Corollary 7.4 *Let ϕ be any first-order sentence about graphs (i.e. quantification over nodes, relations $E(u, v)$ + identity). Then either $G \models \phi$ or $G \models \neg\phi$ for a.e. $G \in \mathcal{G}(n, p)$.*

Proof. Induction on the structure of ϕ , using Lemma 7.2 to eliminate quantifiers.
□

Thus, all the first-order properties of $\mathcal{G}(n, p)$ for fixed p are easily captured. Things are more interesting when the number of nodes discussed and/or the probability p depends on n .

Given graph G , denote:

independence number $\alpha(G)$ = size of the largest independent set in G ,
 clique number $\omega(G)$ = size of the largest clique in G ,
 chromatic number $\chi(G)$ = smallest number of colours needed for colouring nodes in G so that no two adjacent nodes get the same colour.

Lemma 7.5 *Given $n \geq k \geq 2$, random $G \in \mathcal{G}(n, p)$:*

$$\Pr(\alpha(G) \geq k) \leq \binom{n}{k} q^{\binom{k}{2}}.$$

Proof. Probability that given k -set of nodes in G is independent is $q^{\binom{k}{2}}$. Total number of k -sets is $\binom{n}{k}$. □

Theorem 7.6 *Let $p, 0 < p < 1$ and $\varepsilon > 0$ be constant. Then for a.e. $G \in \mathcal{G}(n, p)$:*

$$\chi(G) > \frac{\ln(1/q)}{2 + \varepsilon} \cdot \frac{n}{\ln(n)} = \Omega\left(\frac{n}{\ln(n)}\right) = \text{large!}$$

Proof. By Lemma 7.5, for any fixed $n \geq k \geq 2$:

$$\begin{aligned} \Pr(\alpha(G) \geq k) &\leq \binom{n}{k} q^{\binom{k}{2}} \leq n^k q^{\binom{k}{2}} \\ &= q^{k \frac{\ln n}{\ln q} + \frac{1}{2} k(k-1)} \\ &= q^{\frac{k}{2} [-\frac{2 \ln(n)}{\ln(1/q)} + k - 1]} \\ &\rightarrow 0 \text{ for } k \text{ large,} \end{aligned}$$

when

$$\frac{k}{2} \left[-\frac{2 \ln(n)}{\ln(1/q)} + k - 1 \right] \rightarrow \infty \text{ for } k > (2 + \varepsilon) \frac{\ln(n)}{\ln(1/q)}.$$

Thus there cannot be a.s. k nodes coloured with same colour for large k
 \Rightarrow More than $\frac{n}{k} = \frac{\ln 1/q}{2+\varepsilon} \cdot \frac{n}{\ln n}$ colours needed. \square

Theorem 7.7 Let p , $0 < p < 1$ be constant. Then for a.e. $G \in \mathcal{G}(n, p)$:

$$\omega(G) \in \{d, d+1\},$$

where $d = d(n, p)$ is the largest integer such that

$$\binom{n}{d} p^{\binom{d}{2}} \geq \ln n.$$

(This implies $d = 2 \log_{1/p}(n) + O(\log \log(n))$.) \square

A graph property Q is an isomorphism-closed family of graphs, i.e. if $G \in Q$ (or “ G has Q ”) and $G \approx G'$, then also $G' \in Q$.

A graph property is *monotone* if it is preserved under addition of edges, i.e. if $G = (V, E)$ and $G' = (V, E')$ are graphs such that $E \subseteq E'$ and G has Q , then also G' has Q .

A *threshold function* for the graph property Q is a function $t : \mathbb{N} \rightarrow \mathbb{R}$ such that

$$\Pr(G \in \mathcal{G}(n, p(n)) \text{ has } Q) \xrightarrow{n \rightarrow \infty} \begin{cases} 1, & \text{if } p \succ t \\ 0, & \text{if } p \prec t. \end{cases}$$

Notation:

$$p \succ t \Leftrightarrow \lim_{n \rightarrow \infty} \frac{p(n)}{t(n)} = \infty,$$

$$p \prec t \Leftrightarrow \lim_{n \rightarrow \infty} \frac{p(n)}{t(n)} = 0,$$

$$p \sim t \Leftrightarrow \lim_{n \rightarrow \infty} \frac{p(n)}{t(n)} = 1,$$

$$p \approx t \Leftrightarrow p(n) = \Theta(t(n)).$$

Denote: $P_n^Q(p) = \Pr(G \in \mathcal{G}(n, p) \text{ has } Q)$.

Then for monotone Q : $p_1 \leq p_2 \Rightarrow P_n^Q(p_1) \leq P_n^Q(p_2) \forall n$.

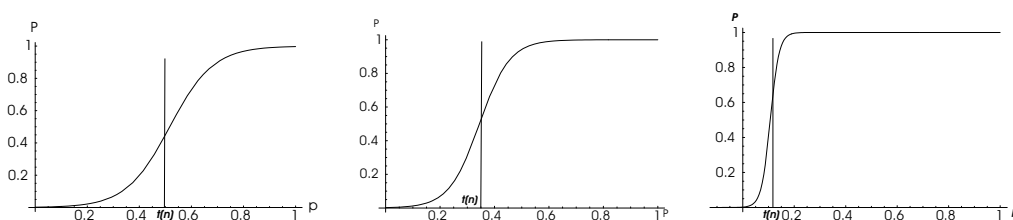


Figure 7: $P_n^Q(p)$ for (a) small, (b) intermediate and (c) large n .

Denote: $p_n^Q(\alpha) =$ the smallest p such that $P_n^Q(p) \geq \alpha$.

In fact $P_n^Q(p)$ is a continuous, strictly increasing function, so really $p_n^Q(\alpha) =$ unique p such that $P_n^Q(p) = \alpha$.

Figure 7 illustrates the evolution of the function P_n^Q , and a corresponding threshold function $t(n)$, for a monotone graph property Q from small to large values of n .

Lemma 7.8 A function $t(n)$ is a threshold for monotone graph property Q if and only if

$$t(n) \approx p_n^Q(\alpha)$$

for all $0 < \alpha < 1$.

Proof.

“ \Rightarrow ” Assume $t(n)$ is a threshold for Q . This means that if $p(n)/t(n) \rightarrow \infty$, then

$$P_n^Q(p(n)) \rightarrow 1 \quad (*)$$

and if $p(n)/t(n) \rightarrow 0$, then

$$P_n^Q(p(n)) \rightarrow 0 \quad (**)$$

Suppose then that

$$t(n) \not\approx p_n^Q(\alpha)$$

for some $0 < \alpha < 1$. This means that either there is a sequence n_1, n_2, \dots such that

$$p_{n_k}^Q(\alpha)/t(n_k) \rightarrow \infty,$$

contradicting (*), or there is a sequence n_1, n_2, \dots such that

$$p_{n_k}^Q(\alpha)/t(n_k) \rightarrow 0,$$

contradicting (**).

(Note that by definition, $P_n^Q(p_n^Q(\alpha)) = \alpha$!)

“ \Leftarrow ” Assume that $t(n)$ is *not* a threshold for Q . Then there is either a sequence n_1, n_2, \dots such that

$$p(n_k)/t(n_k) \rightarrow \infty,$$

but

$$P_n^Q(p(n_k)) \leq \alpha < 1$$

a.e., or a sequence n_1, n_2, \dots such that

$$p(n_k)/t(n_k) \rightarrow 0,$$

but

$$P_n^Q(p(n_k)) \geq \alpha > 0$$

a.e. In the former case,

$$t(n_k) \prec p(n_k) \leq p_{n_k}^Q(\alpha),$$

and in the latter case

$$t(n_k) \succ p(n_k) \geq p_{n_k}^Q(\alpha).$$

Thus in either case $t(n) \not\approx p_n^Q(\alpha)$. \square

Theorem 7.9 *Every monotone graph property Q has a threshold function.*

Proof. For brevity, denote $p_n^Q(\alpha) = p(\alpha)$. Choose some arbitrary $0 < \alpha < \frac{1}{2}$. The goal is to prove that $p(\alpha) \approx p(1 - \alpha)$, thus establishing e.g.

$$t(n) = p\left(\frac{1}{2}\right) = p_n^Q\left(\frac{1}{2}\right)$$

as a threshold function for Q . (Since $p(\alpha) \leq p(\frac{1}{2}) \leq p(1 - \alpha)$.)

Let $m \in \mathbb{N}$ be such that $(1 - \alpha)^m \leq \alpha$. Let $p = p_n(\alpha)$ and consider a sample of m independent graphs G_1, \dots, G_m from $\mathcal{G}(n, p)$. Then the graph $G_1 \cup \dots \cup G_m \in \mathcal{G}(n, q)$, where $q = 1 - (1 - p)^m \leq mp$, and so

$$\Pr(G_1 \cup \dots \cup G_m \text{ has } Q) \leq \Pr(G \in \mathcal{G}(n, mp_n(\alpha)) \text{ has } Q).$$

On the other hand, since Q is monotone, if any G_i has Q , then so does $G_1 \cup \dots \cup G_m$. Thus,

$$\begin{aligned} \Pr(G_1 \cup \dots \cup G_m \text{ does not have } Q) &\leq (1 - \Pr(G_i \text{ has } Q))^m \\ &= (1 - \alpha)^m \leq \alpha. \end{aligned}$$

Hence,

$$\Pr_n^Q(mp_n(\alpha)) \geq \Pr(G_a \cup \dots \cup G_m \text{ has } Q) \geq 1 - \alpha,$$

and so

$$p_n(\alpha) \leq p_n(1 - \alpha) \leq mp_n(\alpha),$$

i.e. $p(\alpha) \approx p(1 - \alpha)$. (Since m depends only on α , not on n .) \square

Consider a graph property Q defined as “ G has Q ” if $X(G) > 0$, where $X \geq 0$ is a random variable on $\mathcal{G}(n, p)$.

E.g. if $X(G)$ denotes the number of spanning trees of G , then property Q corresponds to connectedness.

A threshold function for property Q is a $t(n)$ such that

- (i) $p(n) \prec t(n) \Rightarrow$ almost no $G \in \mathcal{G}(n, p(n))$ has Q .
- (ii) $p(n) \succ t(n) \Rightarrow$ almost all $G \in \mathcal{G}(n, p(n))$ have Q .

If X is integral, then condition (i) can be verified by upper bounding $E[X]$; by Markov’s inequality:

$$\Pr(X \geq 1) \leq E[X] \quad (\text{more generally, for } a > 0 \\ p(X \geq a) \leq E[X]/a).$$

Condition (ii) is trickier, but can be approached by lower-bounding $E[X]$, and upper-bounding $\text{Var}[X]$. (So called “second-moment method”.)

Denote $\mu = E[X]$, $\sigma^2 = \text{Var}[X] = E[(X - \mu)^2] = E[X^2] - \mu^2$.

Recall Chebyshev’s inequality: for any $\lambda > 0$,

$$\Pr(|X - \mu| \geq \lambda) \leq \frac{\sigma^2}{\lambda^2}.$$

Lemma 7.10 *If $\mu > 0$ for n large, and $\frac{\sigma^2}{\mu^2} \rightarrow 0$ as $n \rightarrow \infty$, then $X(G) > 0$ for a.e. $G \in \mathcal{G}(n, p)$.*

Proof. If $X(G) = 0$, then $|X(G) - \mu| = \mu$. Hence

$$\Pr(X = 0) \leq \Pr(|X - \mu| \geq \mu) \leq \frac{\sigma^2}{\mu^2} \rightarrow 0 \text{ as } n \rightarrow \infty. \quad \square$$

Denote the *density* of a graph G by $\delta(G) = \frac{e(G)}{|G|}$.

Say that a graph G is *balanced* if $\delta(G') \leq \delta(G)$ for all subgraphs G' of G .

Theorem 7.11 *Let H be a balanced graph. Then the graph property “ G has a subgraph isomorphic to H ” has threshold function $n^{-1/\delta(H)}$.*

Proof. Let $X(G)$ = number of H -subgraphs of G . Let $k = |H|$, $l = e(H)$, so $\delta(H) = l/k$. Let us first bound $E[X]$ from above. Let $G \in \mathcal{G}(n, p)$, where $p = \gamma n^{-1/\delta(H)} = \gamma n^{-k/l}$ for some $\gamma = \gamma_n \rightarrow 0$, and denote

$$\mathcal{H} = \{\text{all copies of } H \text{ on vertex-set of } G\}.$$

Then $|\mathcal{H}| = \binom{n}{k} h \leq \binom{n}{k} k! \leq n^k$. Here h is the number of different arrangements of H on a set of k vertices, $h = k!/|\text{Aut}(H)|$. Thus

$$\begin{aligned} E[X] &= \sum_{H' \subseteq \mathcal{H}} \Pr(H' \in G) = |\mathcal{H}| \cdot p^l \\ &\leq n^k p^l = n^k (\gamma n^{-k/l})^l = \gamma^l \xrightarrow{\gamma \rightarrow 0} 0. \end{aligned}$$

Thus if

$$p(n) = \gamma_n \cdot n^{-1/\delta(H)} \prec n^{-1/\delta(H)},$$

then $E[X] \xrightarrow{n \rightarrow \infty} 0$.

By Markov's inequality this means that almost no $G \in \mathcal{G}(n, p)$ contains an H -subgraph for large n .

For the other part, we need to bound from above

$$\frac{\sigma^2}{\mu^2} = \frac{1}{\mu^2} (E[X^2] - \mu^2).$$

Let us try to compute:

$$\begin{aligned} E[X^2] &= \sum_{H', H'' \in \mathcal{H}} \Pr(H' \cup H'' \subseteq G) \\ &= \sum_{H', H'' \in \mathcal{H}} p^{e(H') + e(H'') - e(H' \cap H'')} \\ &\leq \sum_{H', H'' \in \mathcal{H}} p^{2l - i\delta(H)}, \end{aligned}$$

where $i = |H' \cap H''|$. (Note that $\delta(H' \cap H'') \leq \delta(H)$.)

Denote then $\mathcal{H}_i^2 = \{(H', H'') \in \mathcal{H}^2 : |H' \cap H''| = i\}$ and compute separately for each i the sum

$$A_i = \sum_{\mathcal{H}_i^2} \Pr(H' \cup H'' \subseteq G)$$

Case $i = 0$:

$$\begin{aligned} A_0 &= \sum_{\mathcal{H}_0^2} \Pr(H' \cup H'' \subseteq G) \\ &= \sum_{\mathcal{H}_0^2} \Pr(H' \subseteq G) \cdot \Pr(H'' \subseteq G) \quad H', H'' \text{ independent} \\ &\leq \sum_{\mathcal{H}^2} \Pr(H' \subseteq G) \cdot \Pr(H'' \subseteq G) \\ &= \left(\sum_{\mathcal{H}} \Pr(H' \subseteq G) \right)^2 \\ &= \mu^2 \end{aligned}$$

Case $i \geq 1$:

$$\begin{aligned} A_i &= \sum_{\mathcal{H}_i^2} \Pr(H' \cup H'' \subseteq G) \\ &= \sum_{H' \in \mathcal{H}} \sum_{\substack{H'' \in \mathcal{H} \\ |H' \cap H''| = i}} \Pr(H' \cup H'' \subseteq G) \\ &\leq |\mathcal{H}| \cdot \binom{k}{i} \binom{n-k}{k-i} h p^{2l} p^{-il/k} \quad h = \frac{k!}{|\text{Aut}(H)|} \\ &\leq |\mathcal{H}| \cdot c_1 n^{k-i} h p^{2l} (\gamma n^{-k/l})^{-il/k} \\ &= \mu \cdot c_1 n^{k-i} h p^l \gamma^{-il/k} n^i \\ &= \mu \cdot c_1 n^k h p^l \gamma^{-il/k} \\ &= \mu c_2 \underbrace{\binom{n}{k}}_{|\mathcal{H}|} h p^l \gamma^{-il/k} \\ &= \mu^2 \cdot c_2 \gamma^{-il/k} \\ &\leq \mu^2 \cdot c_2 \gamma^{-l/k}. \end{aligned}$$

Thus, denoting $c_3 = kc_2$, we get the estimate

$$\frac{E[X^2]}{\mu^2} = \left(\frac{A_0}{\mu^2} + \frac{\sum_i A_i}{\mu^2} \right) \leq 1 + c_3 \gamma^{-l/k}$$

and hence

$$\frac{\sigma^2}{\mu^2} = \frac{E[X^2] - \mu^2}{\mu^2} \leq c_3 \gamma^{-l/k} \xrightarrow{\gamma \rightarrow \infty} 0.$$

Hence, if $p(n) = \gamma_n n^{-k/l}$ for $\gamma_n \rightarrow \infty$, then by Lemma 7.10 $X(G) > 0$ holds for almost every $G \in \mathcal{G}(n, p)$ for large n . \square

Corollary 7.12 For $k \geq 3$, the property of containing a k -cycle has threshold $t(n) = n^{-1}$. (Note: independent of k) \square

Corollary 7.13 For $k \geq 2$, the property of containing a specific tree structure T on k nodes has threshold function $t(n) = n^{-k/(k-1)}$. \square

Corollary 7.14 For $k \geq 2$, the property of containing a k -clique ($\approx K_k$) has threshold function $t(n) = n^{-2/(k-1)}$. \square

Denote $\delta^*(H) = \max\{\delta(H') \mid H' \text{ is subgraph of } H\}$.

Theorem 4.11' The graph property “ G has a subgraph isomorphic to H ” has threshold function $n^{-1/\delta^*(H)}$. \square

Threshold functions for global graph properties

Also known as “the phase transition”.

The “epochs of evolution”: Consider the structure of random graphs $G \in \mathcal{G}(n, p)$, as $p = p(n)$ increases. The following results can be shown (note that $np =$ average node degree):

0. If $p \prec n^{-2}$, then a.e. G is empty.
1. If $n^{-2} \prec p \prec n^{-1}$, then a.e. G is a forest (a collection of trees).
 - The threshold for the appearance of any k -node tree structure is $p = n^{-k/(k-1)}$.