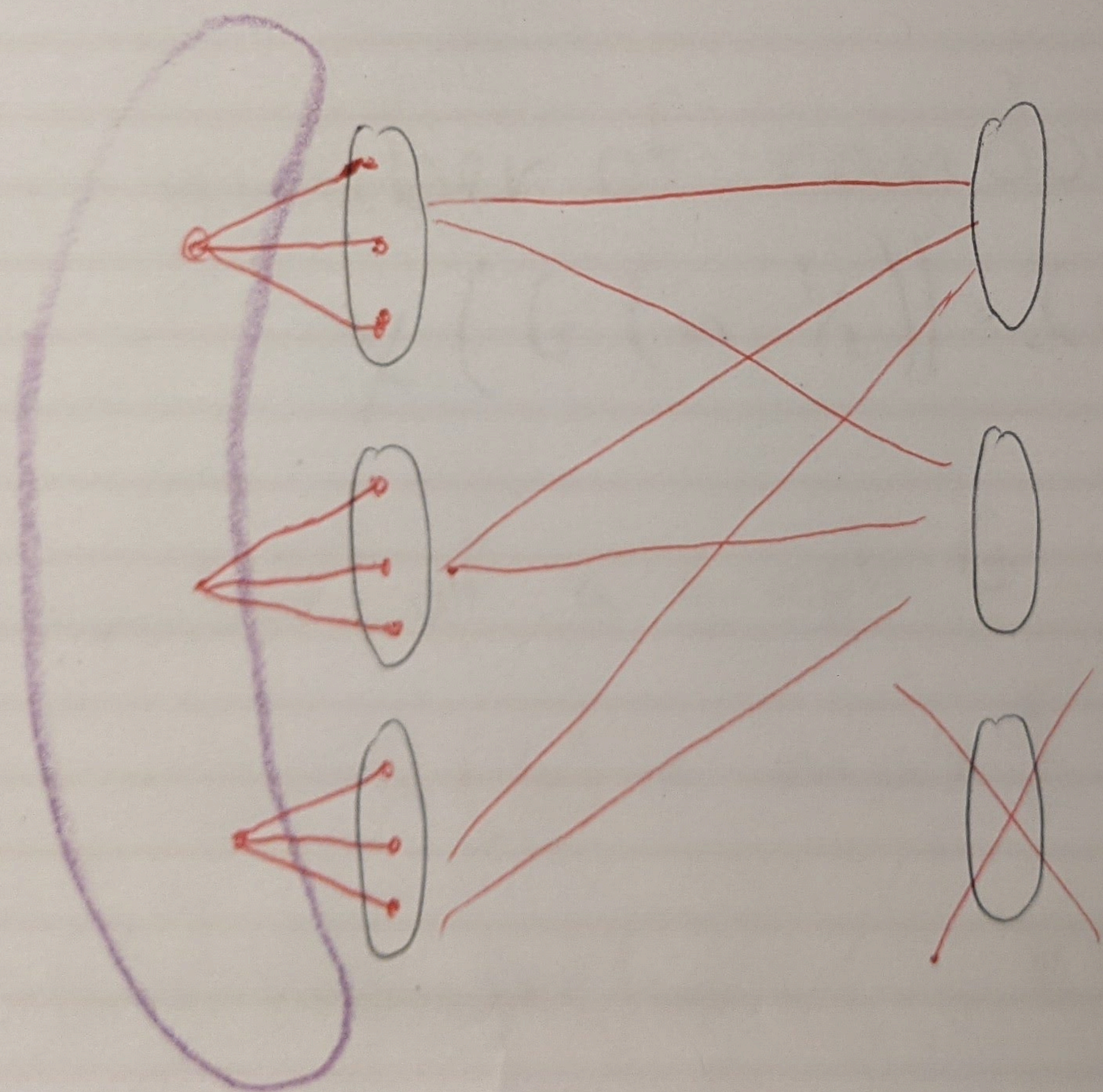


# Lecture 10:

## Graph minors and probabilistic method



A graph  $H$  is a minor of a graph  $G$  if  $H$  can be obtained from a subgraph of  $G$  by contracting edges.

Hadwiger's conjecture:

If  $G$  is a graph with no  $K_{t+1}$  minor then  $G$  is  $t$ -colorable ( $\chi(G) \leq t$ ).

Chromatic number of random graphs:

$$\chi(G(n, \frac{1}{2})) = (1 + o(1)) \frac{n}{2 \log_2 n} \quad \text{w.h.p.}$$

Hadwiger number of a graph  $G$

$$h(G) = \max \{ t : K_t \text{ is a minor of } G \}$$

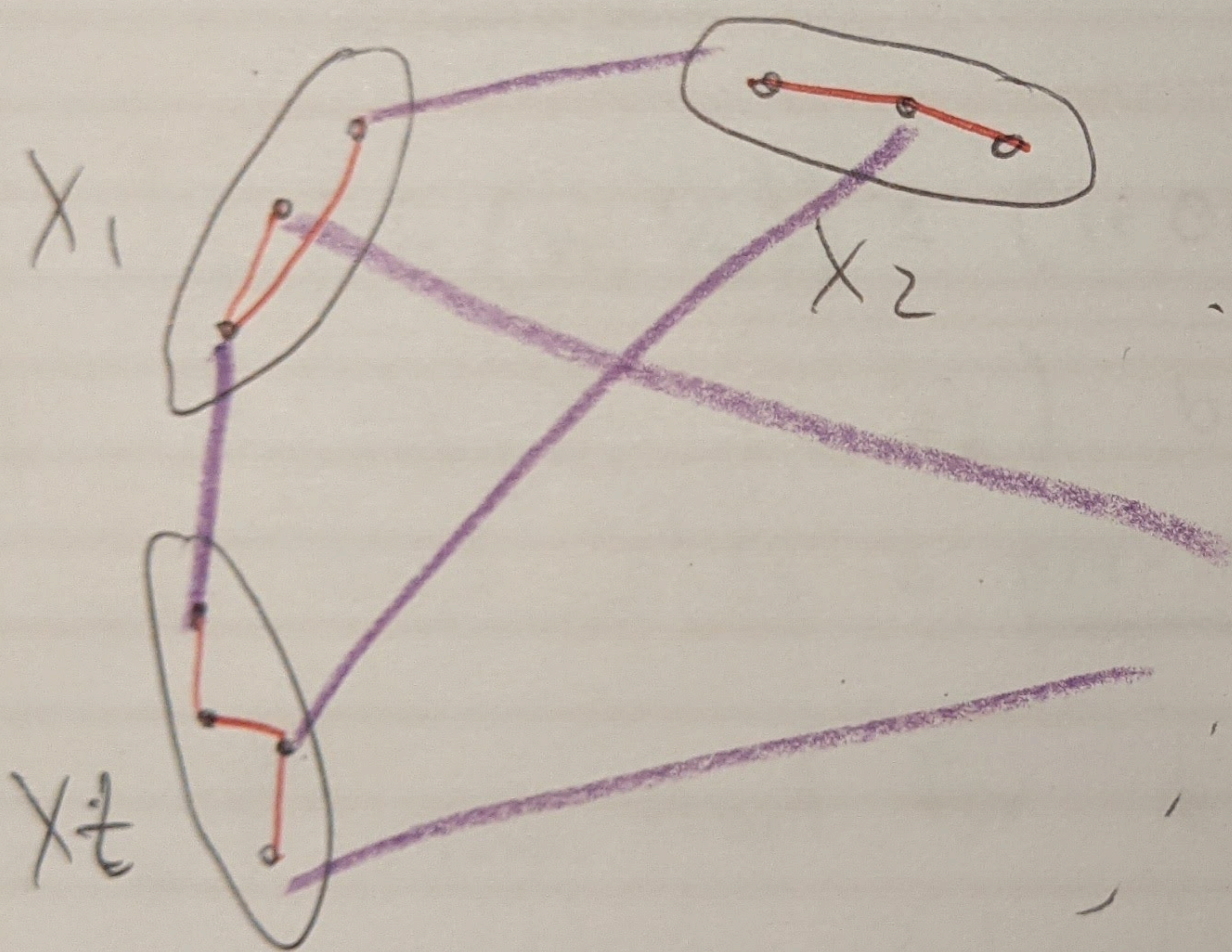
(Hadwiger's conjecture:  $h(G) \geq \chi(G)$ .)

Goal: Determine  $h(G(n, \frac{1}{2}))$ .

A model of  $K_t$  in a graph  $G$  is

a collection  $(X_1, X_2, \dots, X_t)$  of subsets of  $V(G)$ ,  
pairwise disjoint s.t.

- the subgraph of  $G$  induced by  $X_i$  is connected.  
for every  $i$
- for all  $1 \leq i < j \leq t$  there an edge of  $G$   
with one end in  $X_i$  and another in  $X_j$



Observation:

A graph  $G$  contains a model of  $K_t$   
if and only if  $K_t$  is a minor  
of  $G$ .

Proof: Contracting edges within each  
 $X_i$  we can contract it

to a single vertex, the edges  
between  $X_i$ 's yield a complete graph.

Therefore, model  $\Rightarrow$  minor.

In the other direction,  $X_i$ 's  
will correspond to sets contracted

Theorem 5.5:

With high probability

(Bollobás, Catlin, Erdős)

1980

$$h(G(n, 1/2)) = (1 + o(1)) \frac{n}{\sqrt{\log_2 n}}$$

(Hadwiger's conjecture "comfortably" holds for almost every graph.)

Comment:

~~We can~~  $h(G(n, p)) \approx \frac{n}{\sqrt{\log_{1/p} n}}$  for fixed  $p$ .

Proof:

$$h(G(n, 1/2)) \leq (1 + o(1)) \frac{n}{\sqrt{\log_2 n}}$$

Let  $\mathcal{P} = (X_1, X_2, \dots, X_t)$  be a partition of  $[n] = V(G(n, 1/2))$ .

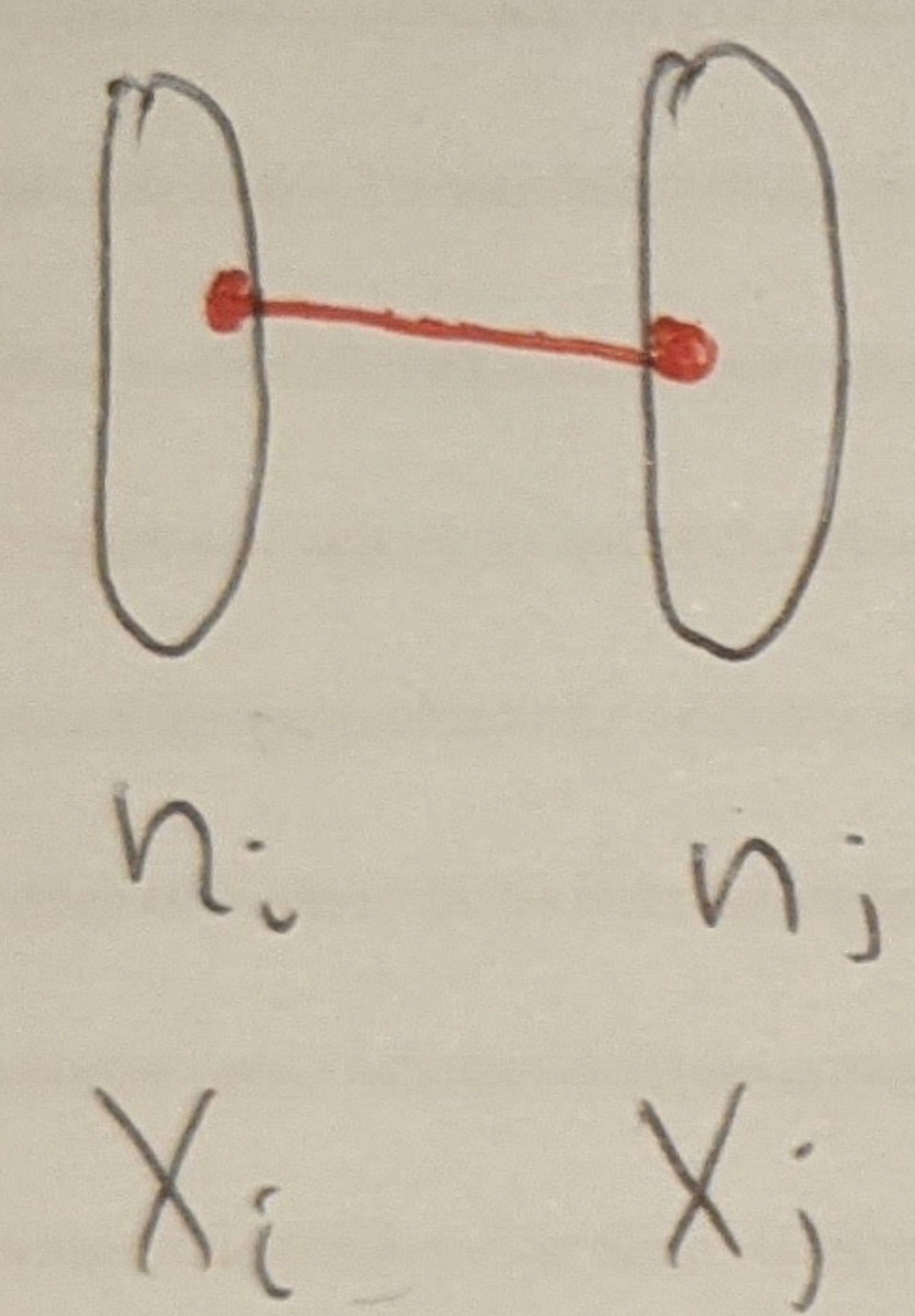
We say that  $\mathcal{P}$  is a pre-model if for all  $i < j$  there is an edge from  $X_i$  to  $X_j$ .

We will show that for  $t = (1 + o(1)) \frac{n}{\sqrt{\log_2 n}}$  w.h.p. there is no pre-model of size  $t$  in  $G(n, 1/2)$  (For the right choice of  $o(1)$  term)

$$\mathbb{P}(\mathcal{P} \text{ is a pre-model}) = \prod_{i < j} (1 - 2^{-n_i n_j}) \leq \prod_{i < j} e^{-2^{-n_i n_j}} = \exp\left(-\sum_{i < j} 2^{-n_i n_j}\right)$$

$$\mathcal{P} = (X_1, X_2, \dots, X_t) \quad |X_i| = n_i$$

$$1 - x \leq e^{-x}$$



$$n_1 + n_2 + \dots + n_t = n$$

$$\frac{\sum_{i < j} 2^{-n_i n_j}}{\binom{t}{2}} \geq \left( \prod_{i < j} 2^{-n_i n_j} \right)^{\frac{2}{\binom{t}{2}}}$$

AM-GM inequality

(?) probability there is an edge from  $X_i$  to  $X_j$

$$2^{-\sum_{i < j} \frac{n_i n_j}{\binom{t}{2}}} \geq 2^{-\frac{\frac{t-1}{t} n^2}{\frac{t(t-1)}{2}}} = 2^{-\frac{n^2}{t^2}}$$

$$2 \sum_{i < j} n_i n_j = \left( \sum n_i \right)^2 - \sum n_i^2 \ll \frac{t-1}{t} n^2$$

$$\mathbb{P}(\text{there exists a pre-model of size } t) \leq t^n e^{-2^{\frac{n^2}{t^2}} \cdot \binom{t}{2}} = o(1)$$

$$t^n \ll e^{-2^{\frac{n^2}{t^2}} \cdot \binom{t}{2}}$$

$$\ln \log t \ll \frac{t^2}{2} 2^{-n^2/4^2}$$

$$n = t (\sqrt{\log_2 t} + \delta)$$

$$n^2/4^2 \approx \log_2 t - \delta \sqrt{\log_2 t}$$

$$\log n \sim \log t$$

$$\ln \log t \ll \frac{t^2}{2} 2^{-\log_2 t + \delta \sqrt{\log_2 t}}$$

$$\ln \log t \ll \frac{t^2}{2} 2^{\delta \sqrt{\log_2 t}}$$

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$$(\log t)^{3/2} \ll 2^{\delta \sqrt{\log_2 t}}$$

$\delta = 1$  already suffices.

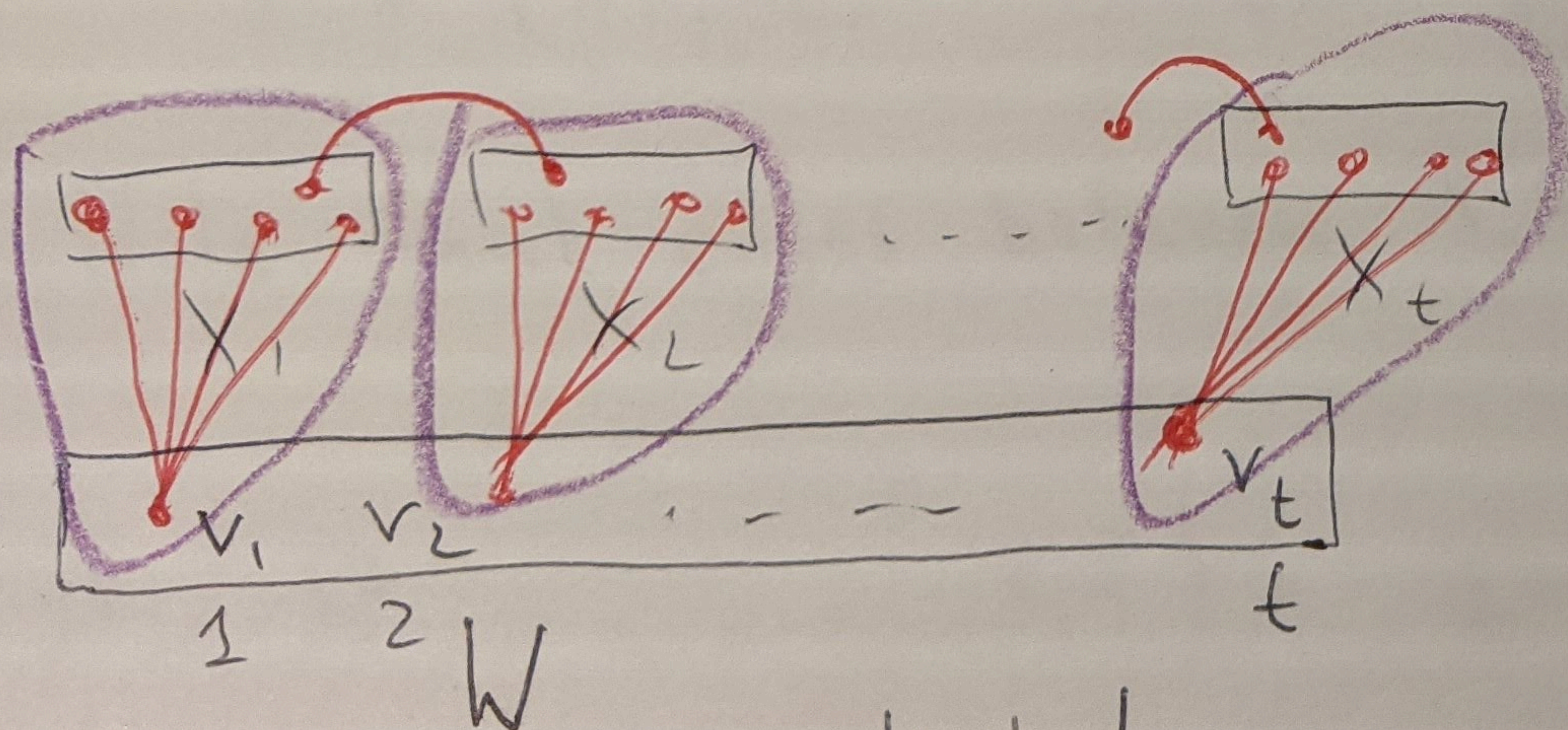
The other direction:

$$h(G(n, 1/2)) \geq (1 - o(1)) \frac{n}{\sqrt{\log_2 n}}$$

$$\text{We prove } \frac{n}{\sqrt{2 \log_2 n}}$$

For every  $\epsilon > 0$  if  $t < (1 - \epsilon) \frac{n}{\sqrt{\log_2 n}}$

and  $n$  is large enough then there is a model of  $K_t$  in  $G(n, 1/2)$



Let  $W = [t] \subseteq V(G(n, 1/2))$  w.h.p.

Let  $k = \sqrt{\log_2 n} + \delta$

$$k = \sqrt{2 \log_2 n} + \delta$$

W.h.p. there exist  $X_1, \dots, X_t \subseteq [n] - W$  pairwise disjoint  
 $|X_i| = k$  s.t. every vertex of  $X_i$   
 is adjacent to  $V_i = \{i\}$ .

Sketch: We select  $X_1, X_2, \dots, X_t$  one by one.

~~After~~ At any step at least  $\frac{\epsilon}{2}n$  vertices  
 are not selected yet, and the previous  
 selection is independent on edges from  
 current vertex  $V_i$ .

By Chernoff bound with probability  
 exponentially close to 1  $V_i$  will have  
 $\geq k$  neighbors not yet selected. which  
 will give  $X_i$ .

It now suffices to show that

$\mathcal{P} = (X_1, X_2, \dots, X_t)$  with high probability  
 Form a premodel of  $K_t$ .

Note that edges between  $X_i$ 's are selected independently of  
 selection of  $X_i$

$$\begin{aligned} \mathbb{P}(\mathcal{P} \text{ is a premodel}) &\geq 1 - \binom{t}{2} \mathbb{P}[X_i \text{ and } X_j \text{ are not joined} \\ &\quad \text{by an edge}] \\ &= 1 - \binom{t}{2} 2^{-k^2} \end{aligned}$$

We need

$$\binom{t}{2} 2^{-k^2} = o(1) \rightarrow \text{this doesn't work}$$
$$\binom{t}{2} 2^{-k^2} \leq \frac{t^2}{2} 2^{-\sqrt{2 \log_2 n}} = \frac{t^2}{2n^2} 2^{-\sqrt{2 \log_2 n}} = o(1)$$

To get  $h(G(n, \frac{1}{2})) \geq (1 - o(1)) \frac{n}{\sqrt{\log_2 n}}$

By the previous calculation the number of edges missing from the premodel

is  $o(t)$ , so by removing  $o(t)$  vertices we get a premodel of  $K_{t - o(t)}$ .



By 5.5. there exist graphs with no  $K_t$  minor  
and average degree  $\sim \frac{t \sqrt{\log_2 t}}{2}$ .

One can do slightly better by considering  $G(n, p)$  for  
 $p \neq \frac{1}{2}$  and optimizing.

getting average degree  $\frac{1 - e^{-2}}{\sqrt{2}} t \sqrt{\log t}$   
for  $p = 1 - e^{-2}$

$$\text{Let } \lambda = \max_{\alpha} \left( \frac{1 - e^{-2\alpha}}{\sqrt{2}} \right)$$

Theorem 5.6: Every graph with average degree  
 $\geq (\lambda + o(1)) t \sqrt{\log t}$   
(Thomason, 2001) contains a  $K_t$  minor.

(This bound is optimum by the above).

Corollary: If  $G$  has no  $K_t$  minor then

$$\chi(G) \leq (\lambda + o(1)) t \sqrt{\log t}$$

Hadwiger's conjecture holds with  $t$  replaced by