

# Some quick notes on exponential random graphs

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# Why?

- ★ we use many random graph models in network applications. . . ■
- ★ but rarely specify the statistical ensemble precisely■
- ★ so even the averages we compute are suspect■
- ★ and even the famous Barabási-Albert scale-free model has known problems■
- ★ it's hard to model degree correlations■

we need a unified, rigorous framework

# Exponential random graphs

- ★ fix a number of nodes  $n$  ■
- ★ consider the set  $G(n)$  of all graphs on  $n$  nodes ■
- ★ we will assign to each  $g \in G(n)$  a probability  $P(g)$  ■
- ★ let  $x = \{x_1, x_2, \dots\}$  be a set of functions on  $G(n)$  representing properties we are interested in, for example
  - ▷  $x_1(g)$ =number of edges
  - ▷  $x_2(g)$ =number of nodes of degree 3
  - ▷  $x_3(g)$ =number of triangles

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- ★ we then assign the probabilities  $P$  by

$$P_{\beta}(g) = \frac{1}{Z(\beta)} \exp(-\beta_1 x_1 - \beta_2 x_2 - \dots)$$

where  $Z(\beta) = \sum_{G(n)} P_{\beta}(g)$  ■

# Statistical mechanics

- ★ Hamiltonian  $H(g) = \sum_i \beta_i x_i(g)$  ■
- ★ free energy:  $F = -\log(Z)$  ■
- ★ entropy  $S = -\sum_{G(n)} P(g) \log(P(g))$  (which is maximized by our choice of  $P$ ) ■
- ★  $\langle x \rangle = \frac{dF}{d\beta} \Big|_{\beta=0}$  ■

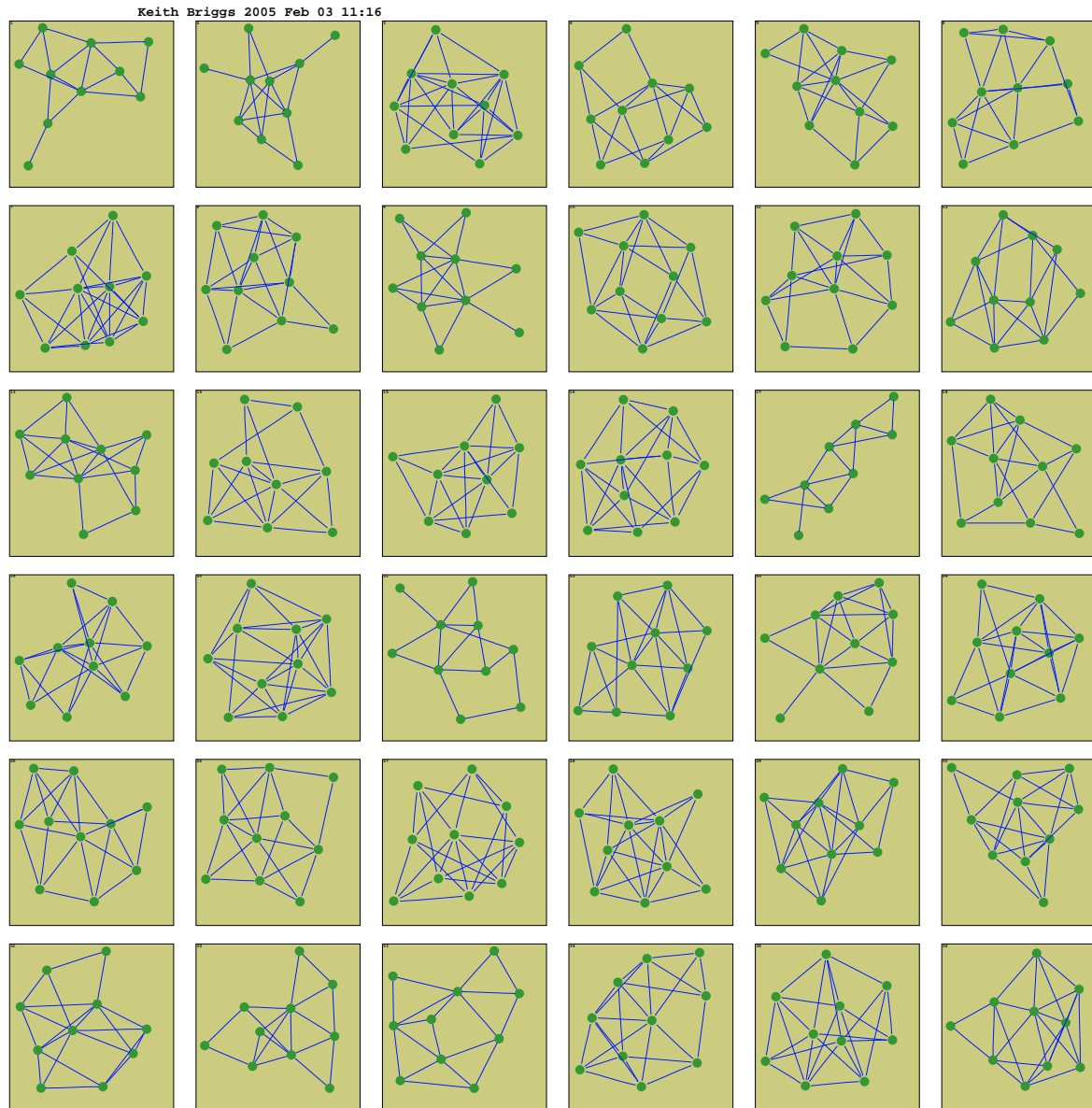
## Exactly soluble example - Bernoulli model $G\{n, p\}$

- ★  $G\{n, p\}$  has  $n$  nodes and each possible edge appears independently with fixed probability  $p$  ■
- ★ let  $x =$  number of edges ■
- ★  $\langle x \rangle = m = \binom{n}{2} p$  ■
- ★  $Z = (1 + \exp(-\beta))^{\binom{n}{2}}$  ■
- ★  $F = \binom{n}{2} \log(1 + \exp(-\beta))$  ■
- ★ which gives  $P(g) = p^m (1-p)^{\binom{n}{2}-m}$  as expected ■
- ★ i.e. each graph appears with equal probability

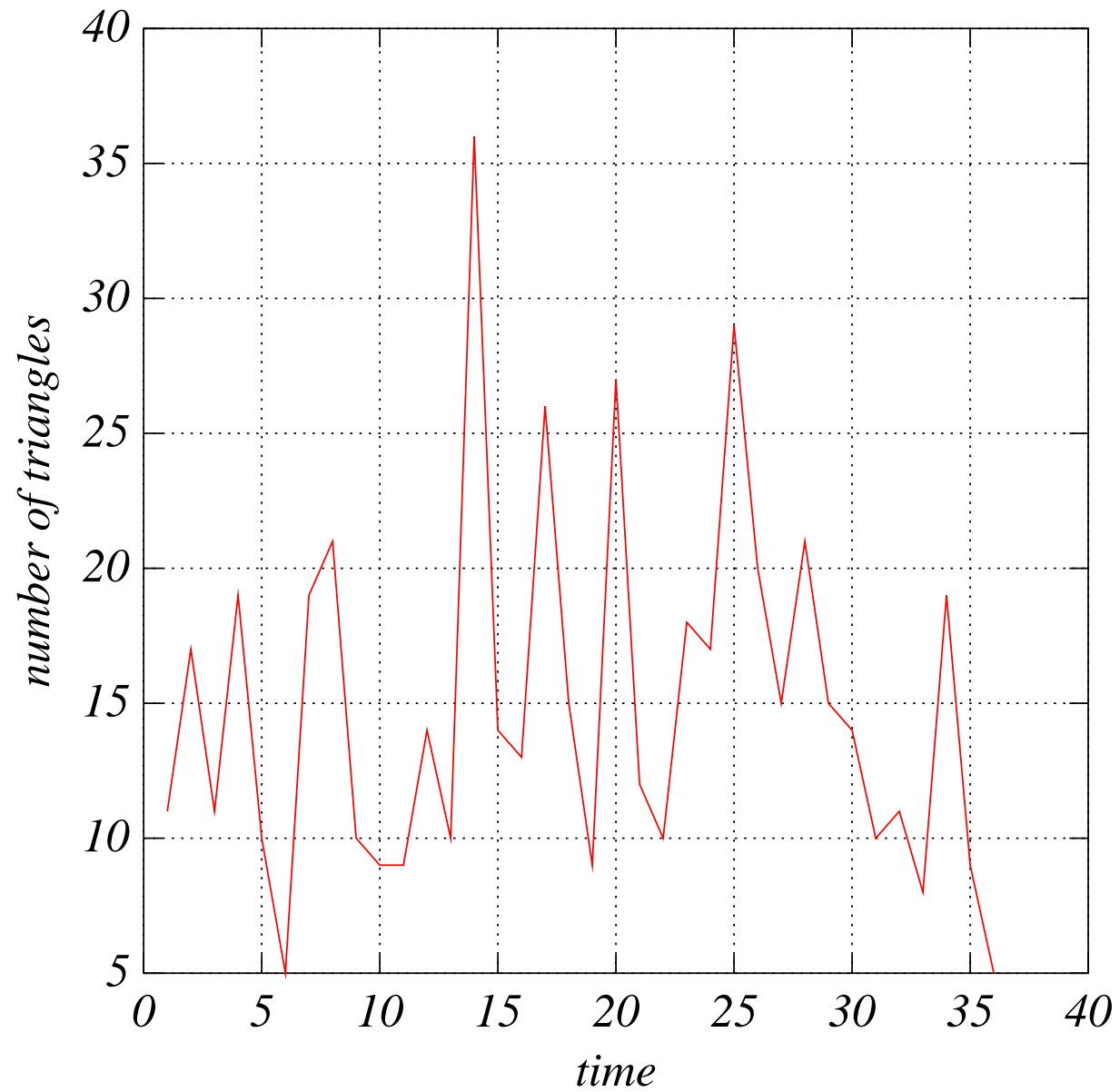
# Metropolis simulation

- ★ works by defining a random walk in  $G$  which has equilibrium distribution equal to our desired  $P$  ■
- ★ typically we choose a pair of nodes, and then flip its state depending on whether the flip is energetically favourable ■

# Simulation example 1 - favouring triangles

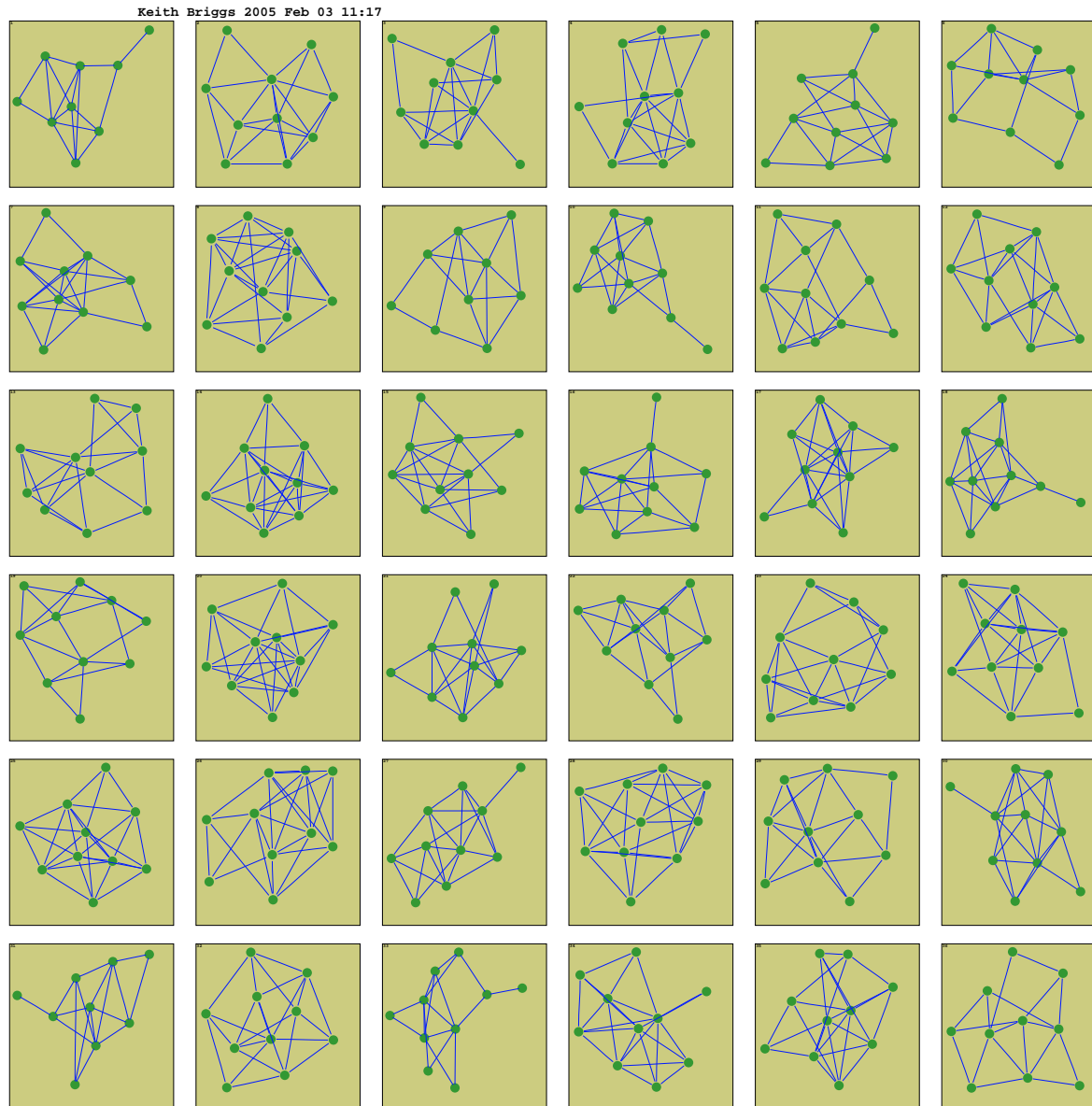


# Simulation example 1 - favouring triangles





# Simulation example 2 - favouring degree 3



## Some references (amongst many)

J. Berg, M. Lässig *Correlated random networks*  
<http://xxx.soton.ac.uk/abs/cond-mat/0205589>

J. Park, M. Newman *The statistical mechanics of networks*  
<http://xxx.soton.ac.uk/abs/cond-mat/0405566>

Z. Burda, J. Correia, A. Krzywicki *Statistical ensemble of scale-free random graphs*  
<http://xxx.soton.ac.uk/abs/cond-mat/0104155>