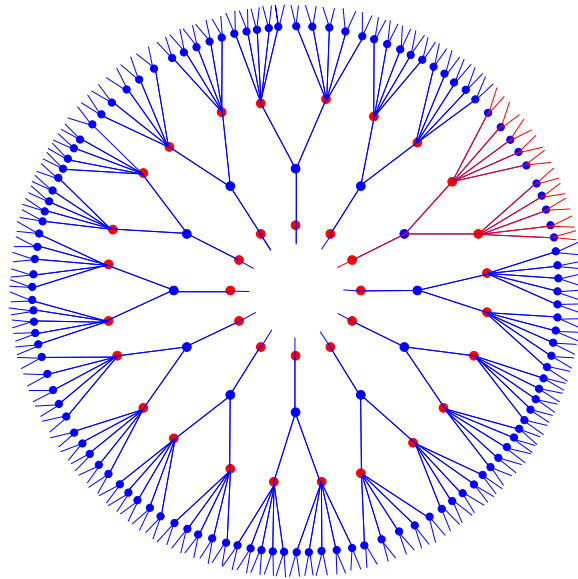


# An Introduction to Low-Density Parity-Check Codes



Amin Shokrollahi

# Outline

We will outline in this talk the **design** and **analysis** of error-correcting codes that can be **encoded** and **decoded** efficiently **and** protect against a fraction of errors that is **almost as large** as given by theoretical upper bounds.

**Existence** of such **bounds** and codes that asymptotically meet these bounds was proved in the landmark paper of **C.E. Shannon** in 1948.

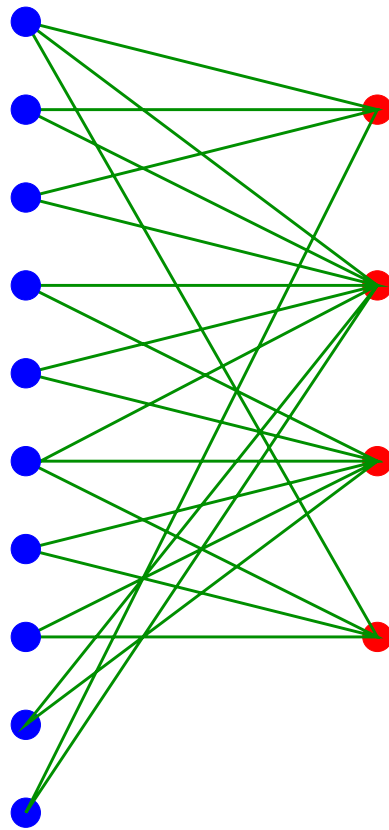
Several codes can be proved to meet the asymptotic bounds. **Almost none of them** are equipped with efficient encoders and decoders.

Gallager	1963
Zyablov	1971
Zyablov-Pinsker	1976
Tanner	1981
Turbo Codes	1993
Berroux-Glavieux-Thitimajshima	

Sipser-Spielman, Spielman	1995
MacKay-Neal, MacKay	1995
Luby-Mitzenmacher-S-Spielman-Stemann	1997
Luby-Mitzenmacher-S-Spielman	1998
Richardson-Urbanke	1999
Richardson-Shokrollahi-Urbanke	1999

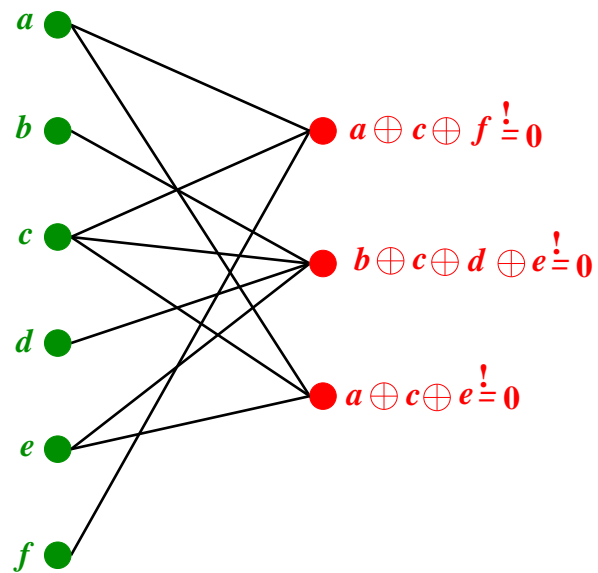
# Code Construction

Codes are constructed from **sparse bipartite graphs**.



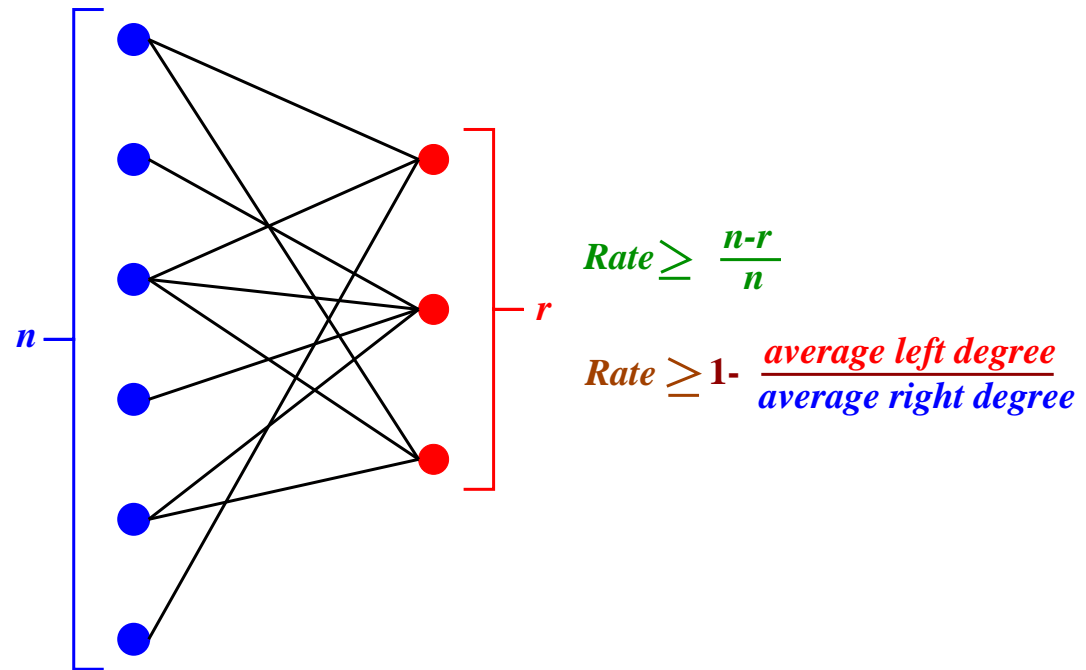
# Code Construction

Any **binary linear code** has a graphical representation.

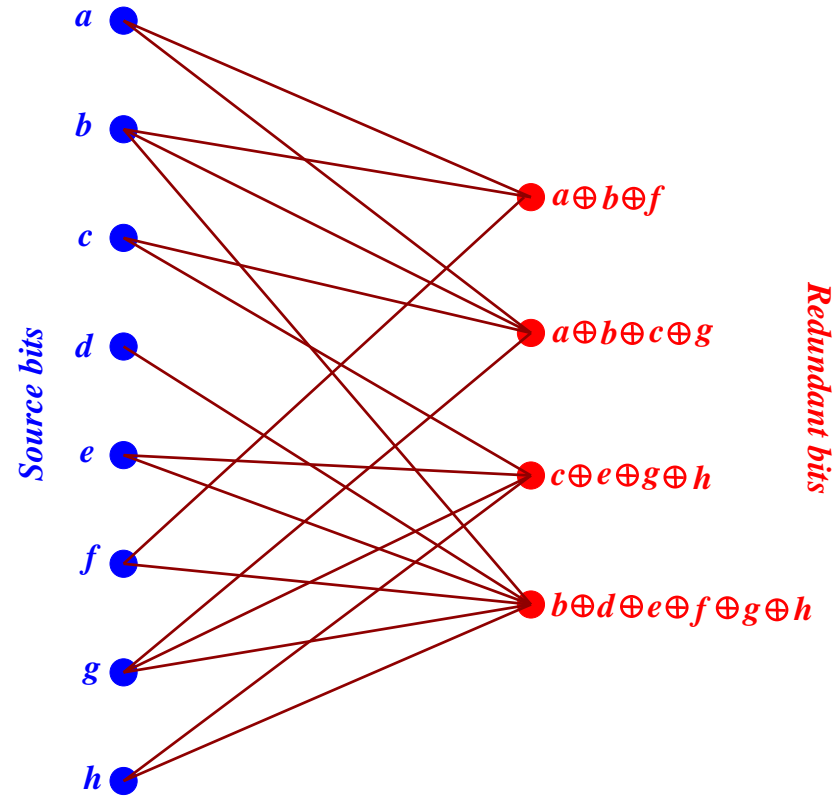


**Not** any code can be represented by a **sparse** graph.

# Parameters



# Dual Construction



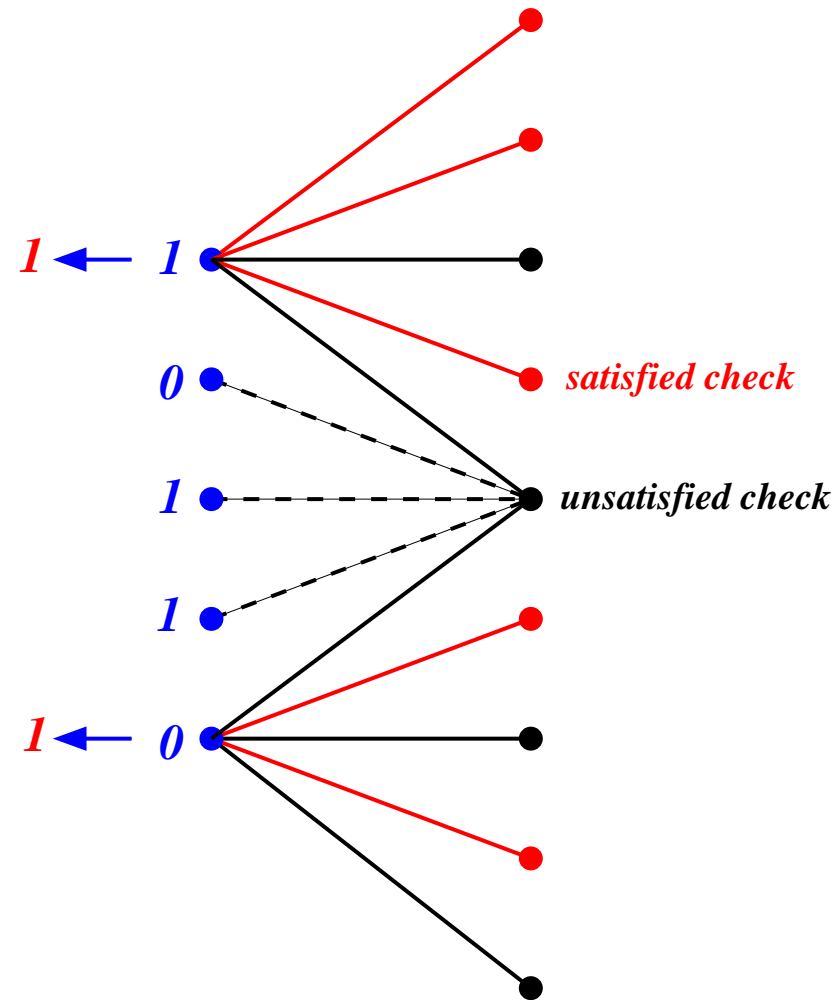
Encoding time is proportional to number of edges.



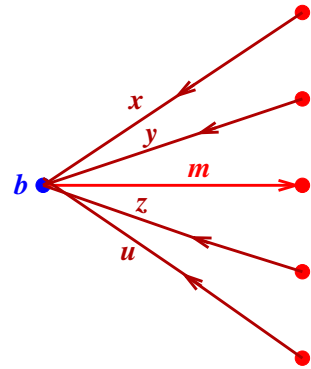
# Algorithmic Issues

- Encoding?
  - Is linear time for the dual construction
  - Is quadratic time (after preprocessing) for the Gallager construction. More later!
- Decoding?
  - Depends on the channel,
  - Depends on the fraction of errors.

# Decoding on a BSC: Flipping

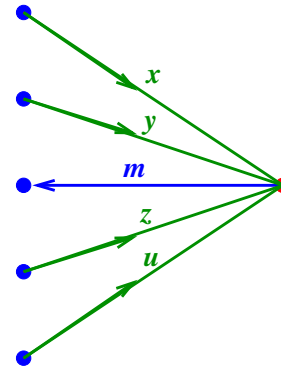


# Decoding on a BSC: Gallager Algorithm A (Message passing)



$$m = \begin{cases} x & \text{if } x=y=z=u \\ b & \text{else} \end{cases}$$

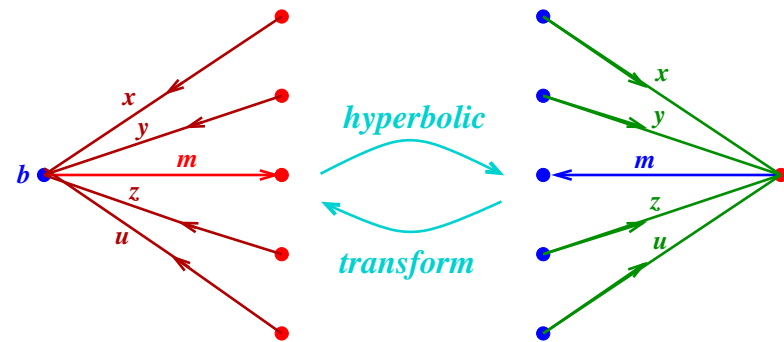
MESSAGE



$$m = x \oplus y \oplus z \oplus u$$

CHECK

# Decoding on a BSC: Belief Propagation



$$m = x + y + z + u + b$$

$$m = x * y * z * u$$

$$(a, b) * (c, d) := (a + c, b + d \bmod 2)$$

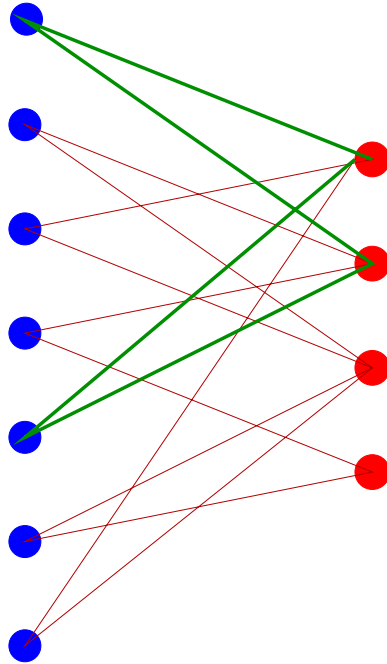
**MESSAGE**

**CHECK**

Messages in **log-likelihood ratios**.

# Optimality of Belief Propagation

Belief propagation is **bit-optimal** if graph has no **loops**.



**Maximizes** the **probability**

$$P(c_m = b | y) = \sum_{c \in C} P(c | y).$$

## Performance on a (3,6)-graph

Shannon limit: 11%

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Flipping algorithm: 1%?

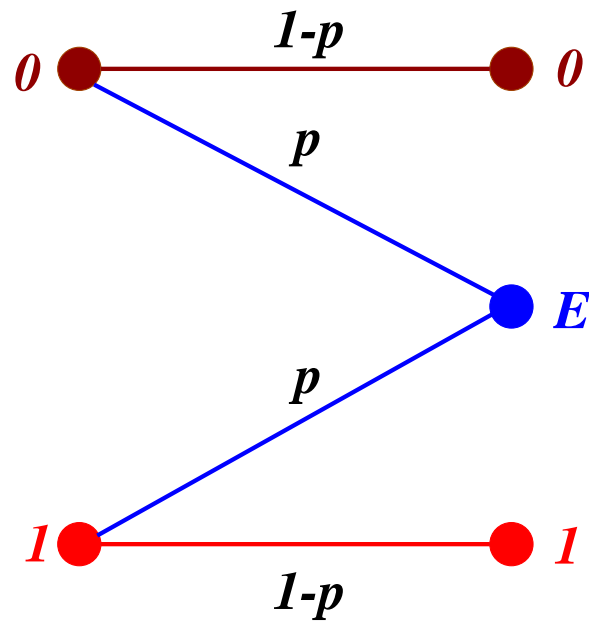
Gallager A: 4%

Gallager B: 4% (6.27%)

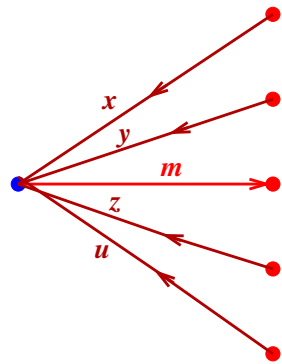
Erasure decoder: 7%

Belief propagation: 8.7% (10.8%)

# The Binary Erasure Channel (BEC)

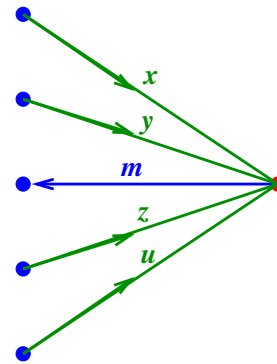


# Decoding on a BEC: Luby-Mitzenmacher-Shokrollahi-Spielman-Stemann



$$m = \begin{cases} 1 & \text{if } x \vee y \vee z \vee u = 1 \\ 0 & \text{else} \end{cases}$$

MESSAGE



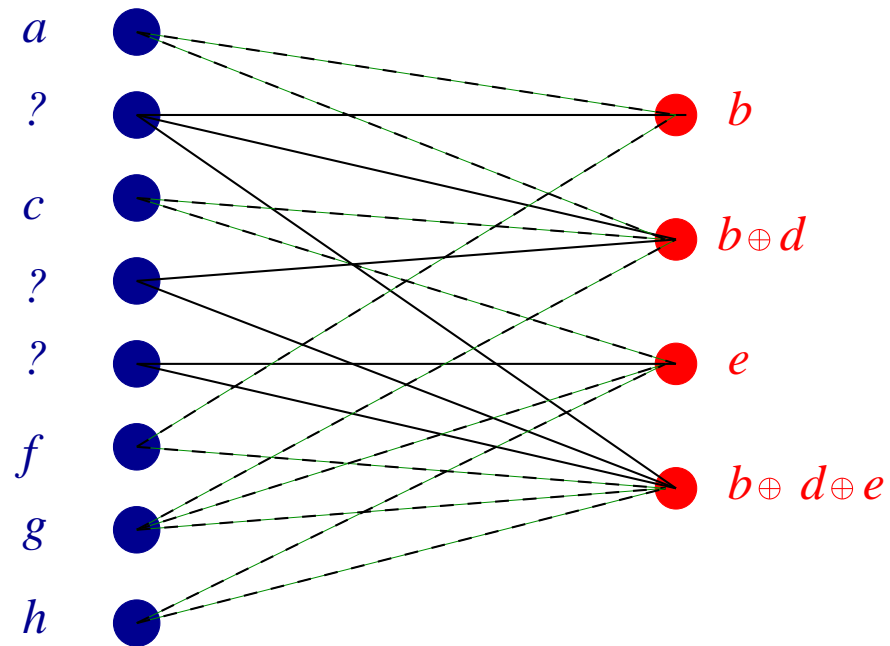
$$m = \begin{cases} 1 & \text{if } x = y = z = u = 1 \\ 0 & \text{else} \end{cases}$$

CHECK



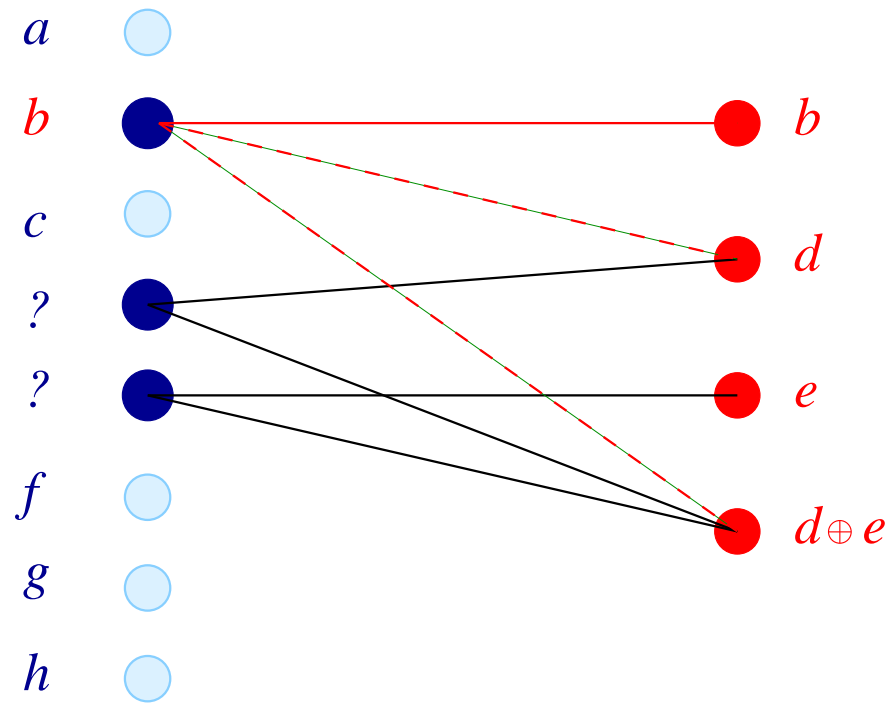
# Decoding on a BEC

Phase 1: Direct recovery

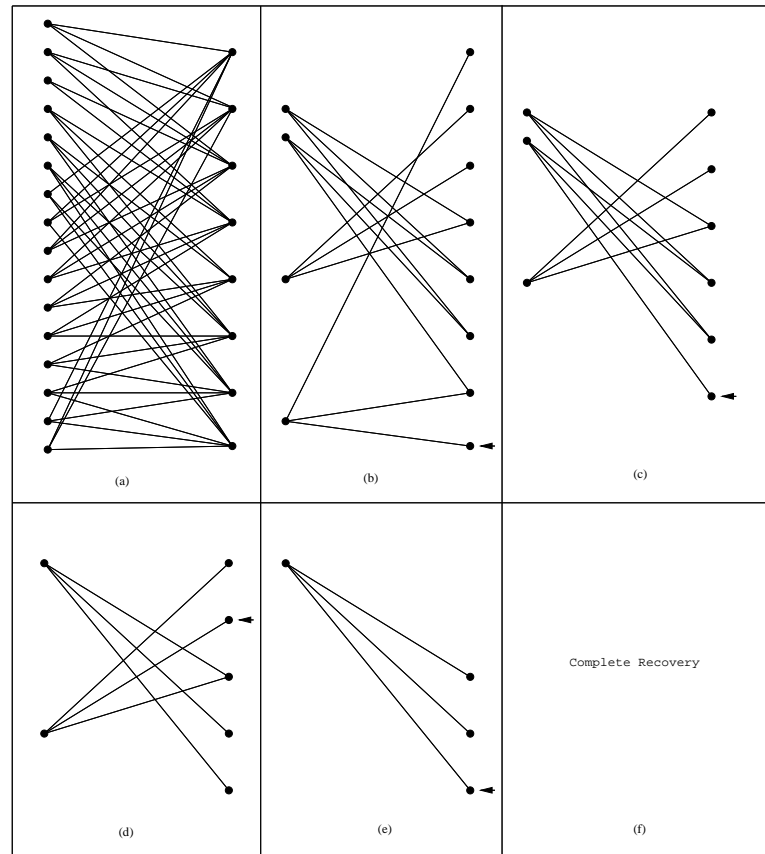


# Decoding on a BEC

Phase 2: Substitution



# Example



## The (inverse) problem

Have: fast decoding algorithms.

Want: design codes that can correct many errors using these algorithms.

Focus on the BEC in the following.

# Experiments

Choose **regular graphs**.

An  $(d, k)$ -regular graph has rate at least  $1 - d/k$ . Can correct **at most** an  $d/k$ -fraction of erasures.

Choose a **random**  $(d, k)$ -graph.

$p_0$  := **maximum** fraction of erasures the algorithm can correct.

$d$	$k$	$d/k$	$p_0$
3	6	0.5	0.429
4	8	0.5	0.383
5	10	0.5	0.341
3	9	0.33	0.282
4	12	0.33	0.2572

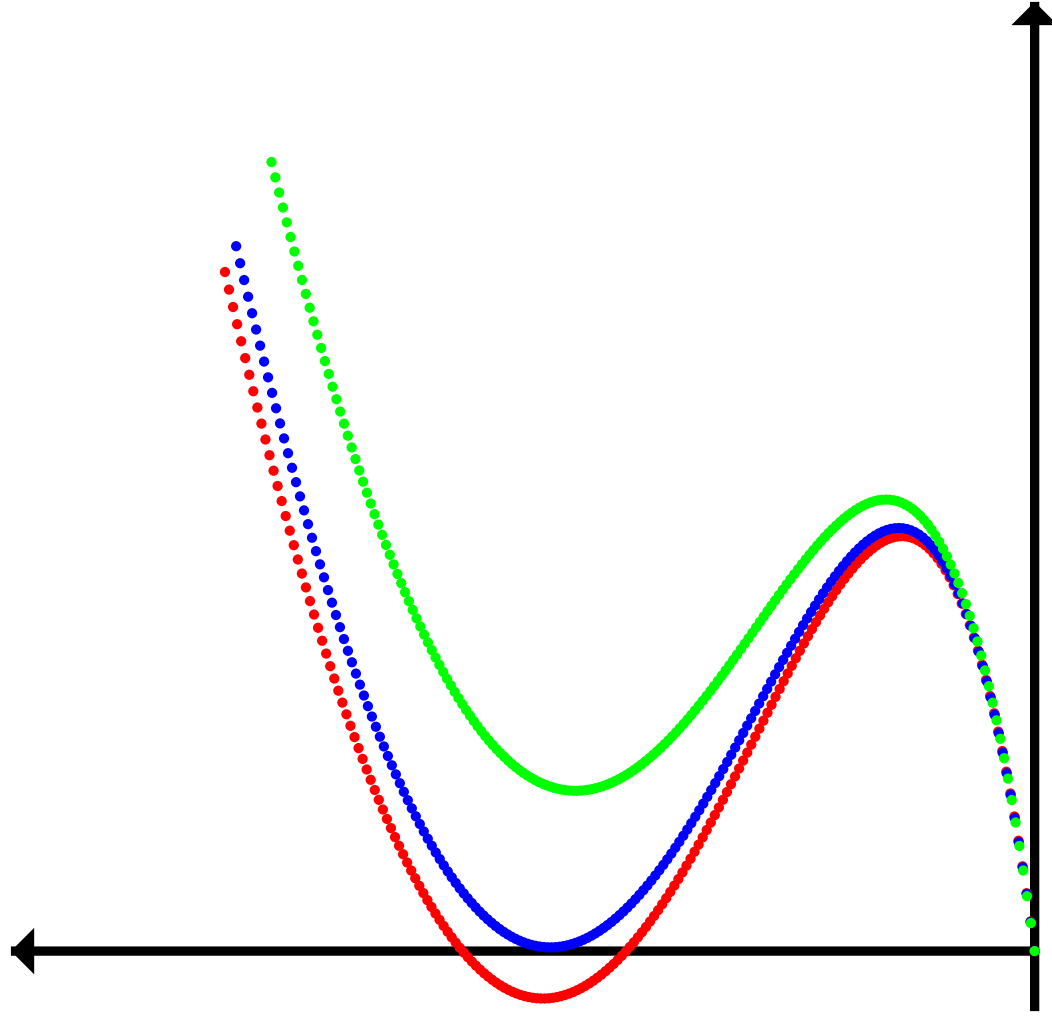
What are these numbers?

## A Theorem

Luby, Mitzenmacher, Shokrollahi, Spielman, Stemmann, 1997:

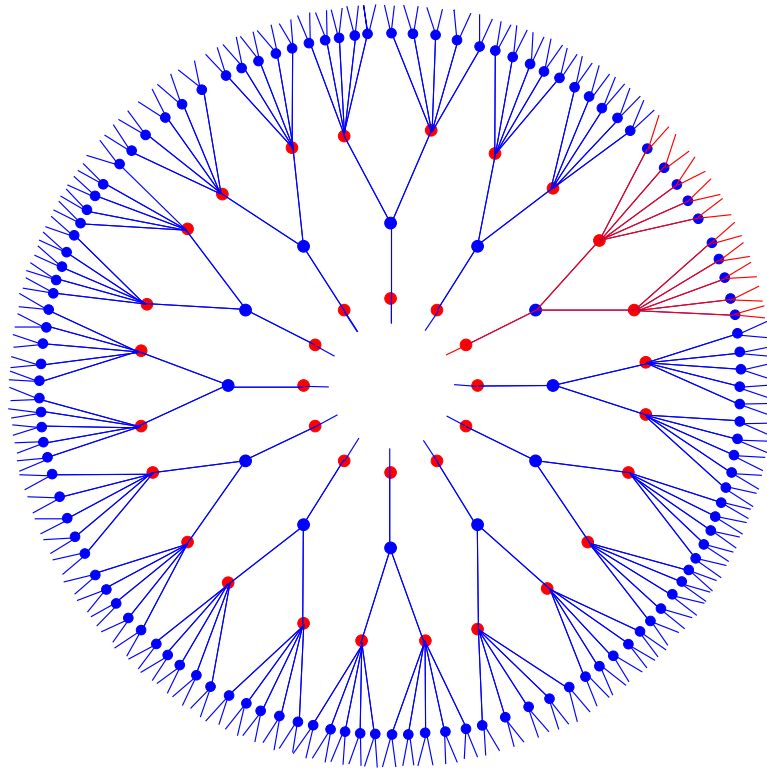
A randomly chosen  $(d, k)$ -graph can correct a  $p_0$ -fraction of erasures with high probability if and only if

$$p_0 \cdot (1 - (1 - x)^{k-1})^{d-1} < x \quad \text{for } x \in (0, p_0).$$



## Analysis: $(3, 6)$ -graphs

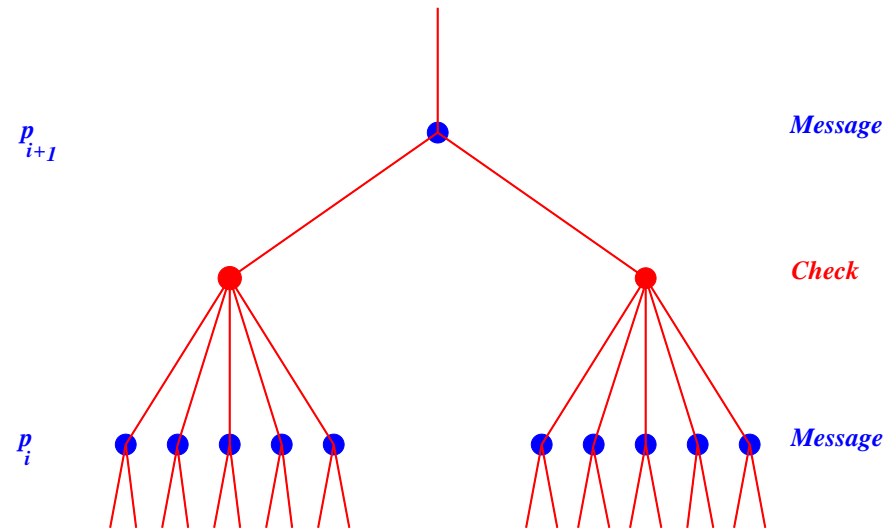
Expand neighborhoods of message nodes.





## Analysis: (3, 6)-graphs

$p_i$  probability that message node is still erased after  $i$ th iteration.



$$p_{i+1} = p_0 (1 - (1 - p_i)^5)^2.$$

# Successful Decoding

Condition:

$$p_0(1 - (1 - p_i)^5)^2 < p_i$$

## Analysis: (3, 6)-graphs

Making arguments **exact**:

- Neighborhood is **tree-like**: **high probability**, standard argument.
- Above argument works for **expected fraction** of erasures at  $\ell$ th round.

Real value is **sharply concentrated** around expected value  $p_\ell$ : **Edge exposure martingale**, Azuma's inequality.

## The General Case

Let  $\lambda_i$  and  $\rho_i$  be the fraction of edges of degree  $i$  on the left and the right hand side, respectively.

Let  $\lambda(x) := \sum_i \lambda_i x^{i-1}$  and  $\rho(x) := \sum_i \rho_i x^{i-1}$ .

Condition for successful decoding for erasure probability  $p_0$  is then

$$p_0 \lambda(1 - \rho(1 - x)) < x$$

for all  $x \in (0, p_0)$ .

## Belief propagation

Richardson-Urbanke, 1999:

$f_\ell$ : density of the probability distribution of the messages passed from the check nodes to the message nodes at round  $\ell$  of the algorithm.

$P_0$ : density of the error distribution (in **log-likelihood representation**).

Consider  $(d, k)$  regular graph.

$$\Gamma(f_{\ell+1}) = \left( \Gamma(P_0 \otimes f_\ell^{\otimes(k-1)}) \right)^{\otimes(d-1)},$$

where  $\Gamma$  is a hyperbolic change of measure function,

$$\Gamma(f)(y) := f(\ln \coth y/2) / \sinh(y),$$

and  $\otimes$  denotes **convolution**.

We want  $f_\ell$  to converge to a **Delta function at  $\infty$** .

Gives rise to **high-dimensional optimization algorithms**.

## Achieving capacity

Want to **design** codes that can recover from a fraction of  $1-R$  of erasures (asymptotically).

Want to have  $\lambda$  and  $\rho$  so that

$$p_0 \lambda(1 - \rho(1 - x)) < x$$

for **all**  $x \in (0, p_0)$ , and  $p_0$  **arbitrarily** close to

$$1 - R = \frac{\int_0^1 \rho(x) dx}{\int_0^1 \lambda(x) dx}.$$

## Tornado codes

Extremely *irregular* graphs provide for *any* rate  $R$  sequences of codes which come arbitrarily close to the capacity of the erasure channel!

Degree structure?

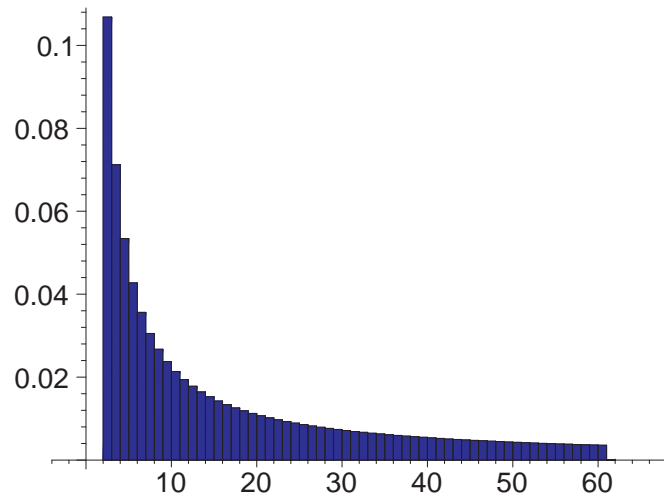
Choose *design parameter*  $D$ .

$$\lambda(x) := \frac{1}{H(D)} \left( x + \frac{x^2}{2} + \cdots + \frac{x^D}{D} \right)$$

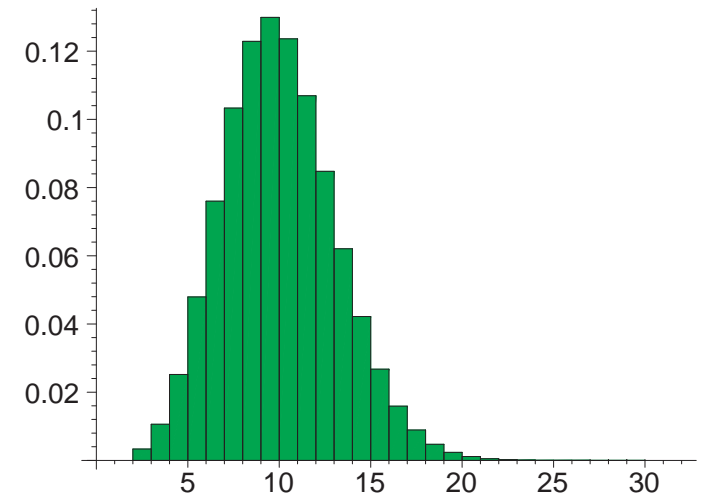
$$\rho(x) := \exp(\mu(x - 1)),$$

where  $H(D) = 1 + 1/2 + \cdots + 1/D$  and  $\mu = H(D)/(1 - 1/(D + 1))$ .

# Tornado Codes: Degree Distribution



Heavy tail



Poisson



## Right regular codes

Shokrollahi, 1999:

Graphs that are **regular** on the right.

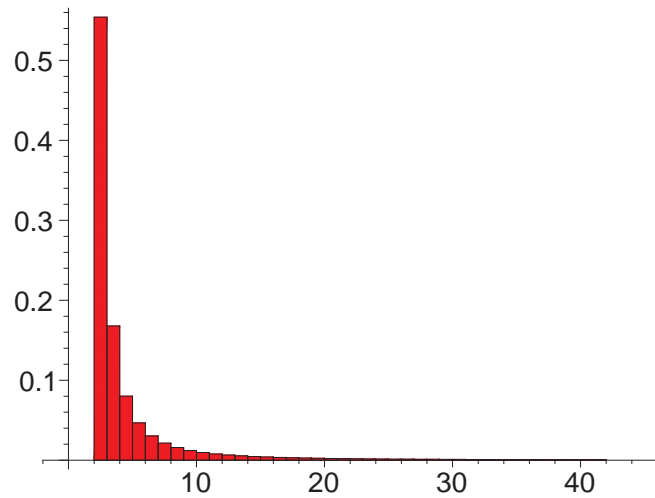
Degrees **on the left** are related to the Taylor expansion of

$$(1 - x)^{1/m}.$$

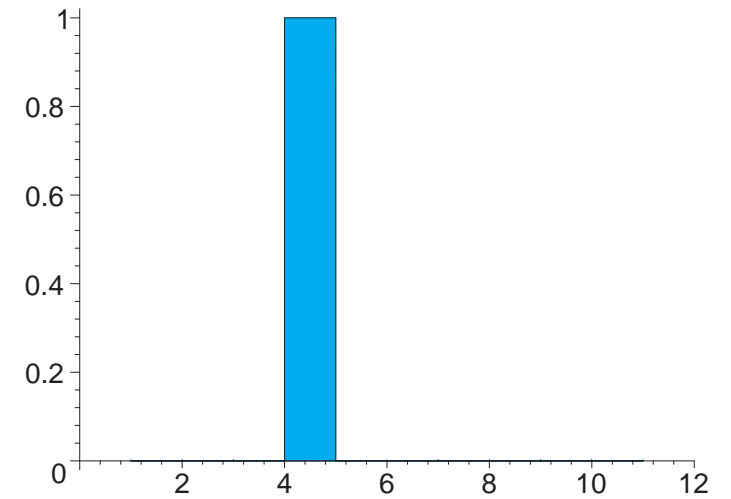
Methodology for constructing capacity-achieving sequences by Oswald-Shokrollahi, 2000.

Also show that the right regular sequence is the **best** in a certain sense.

# Right Regular Codes: Degree Distribution



Left



Right

## Other channels?

$f$  density function.

$$\lambda(f) := \sum_i \lambda_i f^{\otimes(i-1)}.$$

$$\rho(f) := \sum_i \rho_i f^{\otimes(i-1)}.$$

$$\Gamma(f_{\ell+1}) = \rho(\Gamma(P_0 \otimes \lambda(f_\ell))).$$

Want  $P_0$  such that  $f_\ell \rightarrow \Delta_\infty$ .

## Conditions on the density functions

Richardson-Shokrollahi-Urbanke, 1999:

- **Consistency**: if the channel is "*symmetric*", then the density functions  $f_\ell$  satisfy  $f(x) = f(-x)e^x$ .
- **Fixed point theorem**: If  $P_{\text{err}}(f_i) = P_{\text{err}}(f_j)$  for  $i < j$ , then  $f_i = f_j$  is a **fixed point** of the iteration.

## Conditions on the density functions

- **Stability:** let

$$r := - \lim_{n \rightarrow \infty} \frac{1}{n} \log P_{\text{err}}(P_0^{\otimes n}).$$

Then for  $\lambda_2 \rho'(1) > e^r$  we have  $P_{\text{err}}(f_\ell) > \epsilon$  for some fixed  $\epsilon$  and all  $\ell$ .

If  $\lambda_2 \rho'(1) < e^r$ , then the fixed point  $\Delta_\infty$  is **stable**.

$$P_{\text{err}}(f) := \int_{-\infty}^0 f(x) dx$$

is the **error probability**.

## Stability

- Erasure channel with erasure probability  $p_0$ :

$$\lambda_2 \rho'(1) \leq \frac{1}{p_0}.$$

- BSC channel: with probability  $p$ :

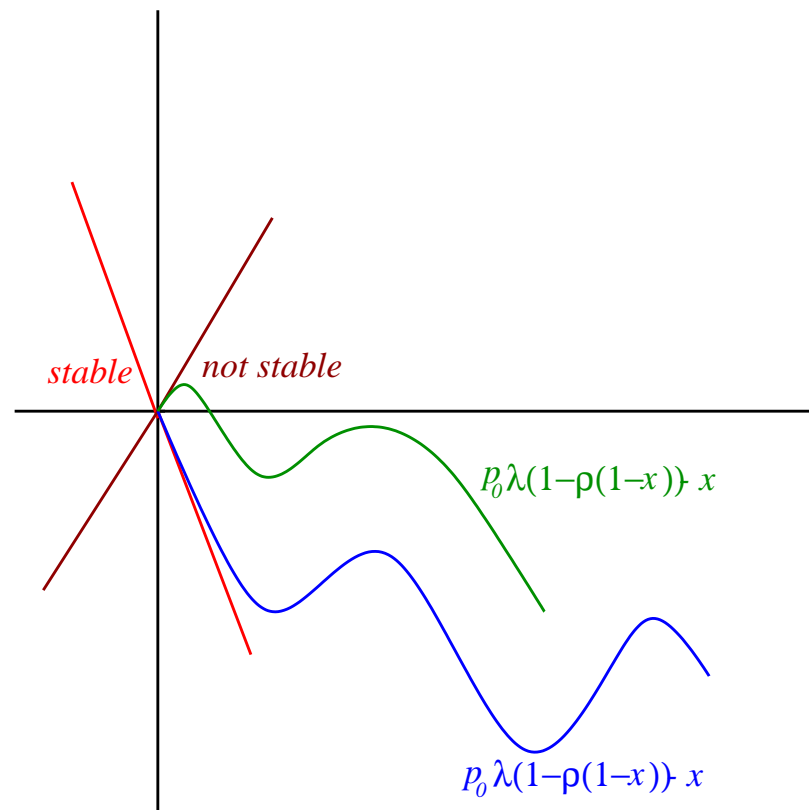
$$\lambda_2 \rho'(1) \leq \frac{1}{2\sqrt{p(1-p)}}.$$

- AWGN channel: with variance  $\sigma^2$ :

$$\lambda_2 \rho'(1) \leq e^{-\frac{1}{2\sigma^2}}.$$

# Stability for the Erasure Channel

Shokrollahi, 1999:



## Flatness: Higher Stability Conditions

Shokrollahi, 2000:

$(\lambda_m(x), \rho_m(x))$  capacity achieving sequence of degree distributions.

Then:

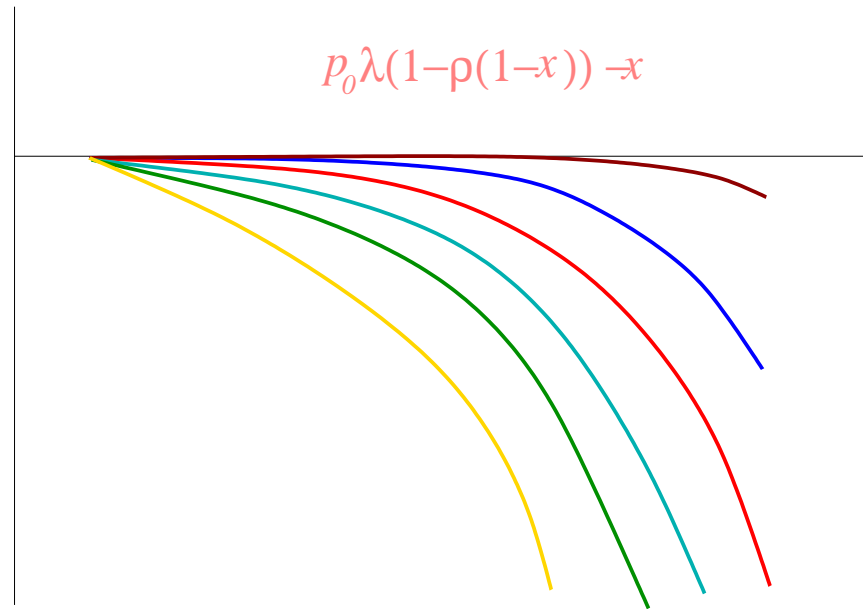
$$(1 - R)\lambda_m(1 - \rho_m(1 - x)) - x$$

converges **uniformly** to the zero-function on the interval  $[0, 1 - R]$ .

**No** equivalent known for other channels.



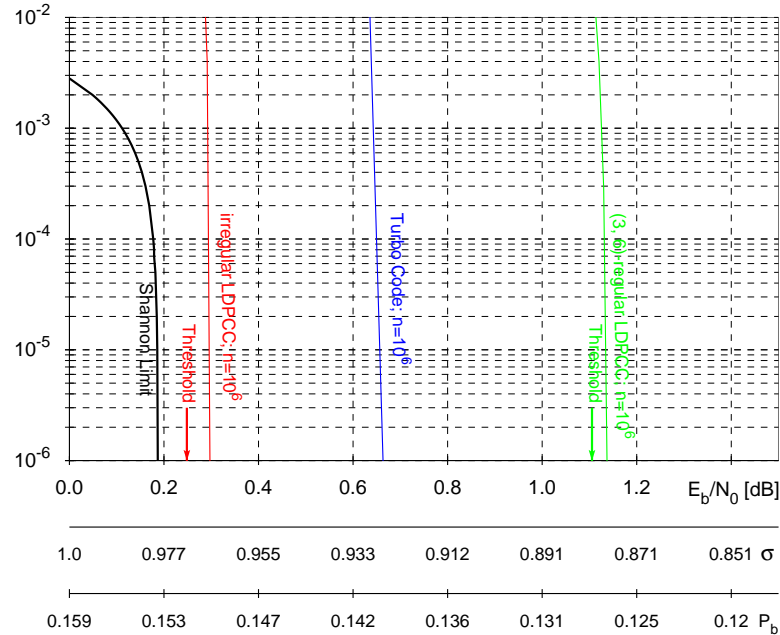
## Flatness: Higher Stability Conditions



# Capacity achieving

No sequences of c.a. degree distributions for channels other than the erasure channel known.

Conjecture: They exist!



# Open problems

## Asymptotic theory

1. **Classification** of capacity achieving sequences for the erasure channel.
2. **Capacity achieving sequences** for **other** channels.
3. **Exponentially small** error probabilities for the decoder (instead of **polynomially small**).

## Explicit constructions

1. Constructions using **finite geometries**.
2. Construction using **Reed-Solomon-Codes**.
3. **Algebraic** constructions.

## Short codes

Graphs with **loops**.

## Algorithmic issues

1. Design and analysis of new **decoding algorithms**.
2. Design of new **encoders**.

### Randomness

Use of **randomness** in other areas: **random convolutional codes?**.