

LP Decoding Achieves Capacity

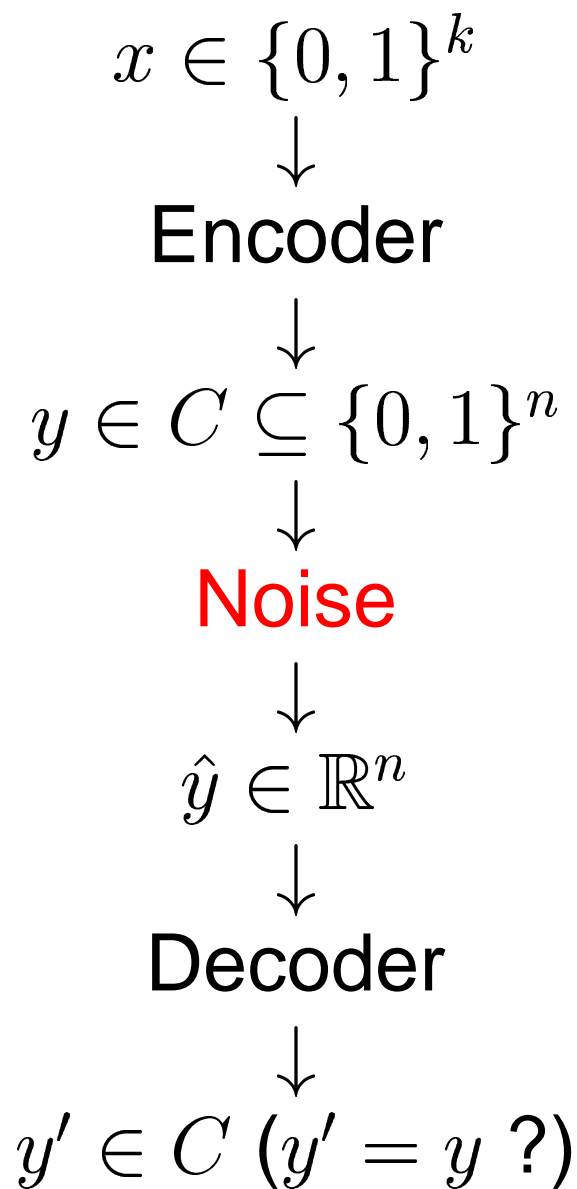
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Thank you: Alexander Barg, David Karger, Ralf Koetter,
Tal Malkin, Rocco Servedio, Pascal Vontobel,
Martin Wainwright, Gilles Zémor.

Codes, Noisy Channels, Achieving Capacity



- Noise: probabilistic (Gaussian).
- Word Error Rate (WER) = $\Pr_{\text{noise}} [y' \neq y]$
- Channel “capacity” \mathcal{C} : highest rate ($r = k/n$) s.t. \exists code family, decoder, with $\text{WER} \leq 2^{-\Omega(n)}$.
- Shannon: characterized capacity for many channels.
- “Achieving capacity:” code family, decoder, with $\text{WER} \leq 2^{-\Omega(n)}$ for all $r < \mathcal{C}$.

Achieving Capacity with Poly-time Decoders

- Forney ('66):
 - ◆ ML/BD decoder
 - ◆ Concatenated codes (OPT \diamond Reed-Solomon)
- Barg/Zémor ('02):
 - ◆ ML/message-passing decoder
 - ◆ Expander codes [SS '96][BZ '01-'04][GI '01-'04]
- This paper:
 - ◆ Linear Programming (LP) decoder
 - ◆ Same expander codes as Barg/Zémor
 - ◆ New feature: “Maximum-Likelihood (ML) Certificate” property: if codeword output, it maximizes $\Pr[\text{correct}]$.

Application to Practical Codes

- Turbo [BGT '93], low-density parity-check (LDPC) [Gal '63] codes:
 - ◆ Practical construction/encoding, moderate length
 - ◆ Perform well (experimentally) under message-passing decoder
- Most successful theory: density evolution [RU, LMSS, RSU, BRU, CFDRU, ..., '99...present].
 - ◆ **Non-constructive, assumes “local tree” structure.**
- LP decoding bounds:
 - ◆ No tree assumption.
 - ◆ Bounds relevant for finite lengths.
 - ◆ Performance of message-passing is closely related [FKW 02, Fel02, KV02, KV04a, KV04b].

Maximum-Likelihood (ML) Decoding

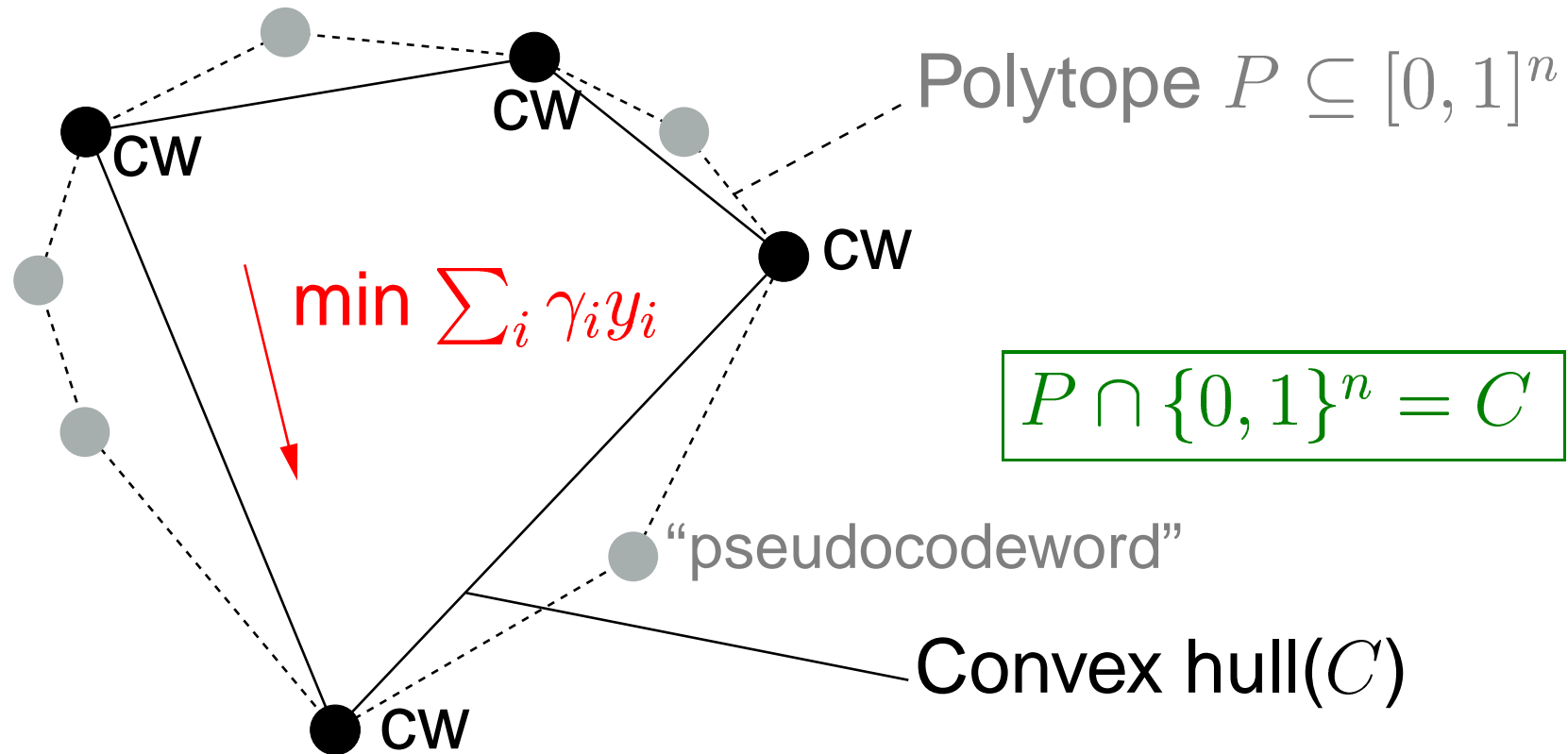
- “Memoryless” channel: each y_i is affected by noise independently.
 - ◆ **Example: Gaussian, where $\hat{y}_i = \mathcal{N}(2y_i - 1, \sigma^2)$.**
- Cost function γ_i : *log-likelihood ratio* of \hat{y}_i .

$$\gamma_i = \ln \left(\frac{\Pr[\hat{y}_i \mid y_i = 0]}{\Pr[\hat{y}_i \mid y_i = 1]} \right)$$

- If y_i more likely 0 $\implies \gamma_i > 0$
- If y_i more likely 1 $\implies \gamma_i < 0$

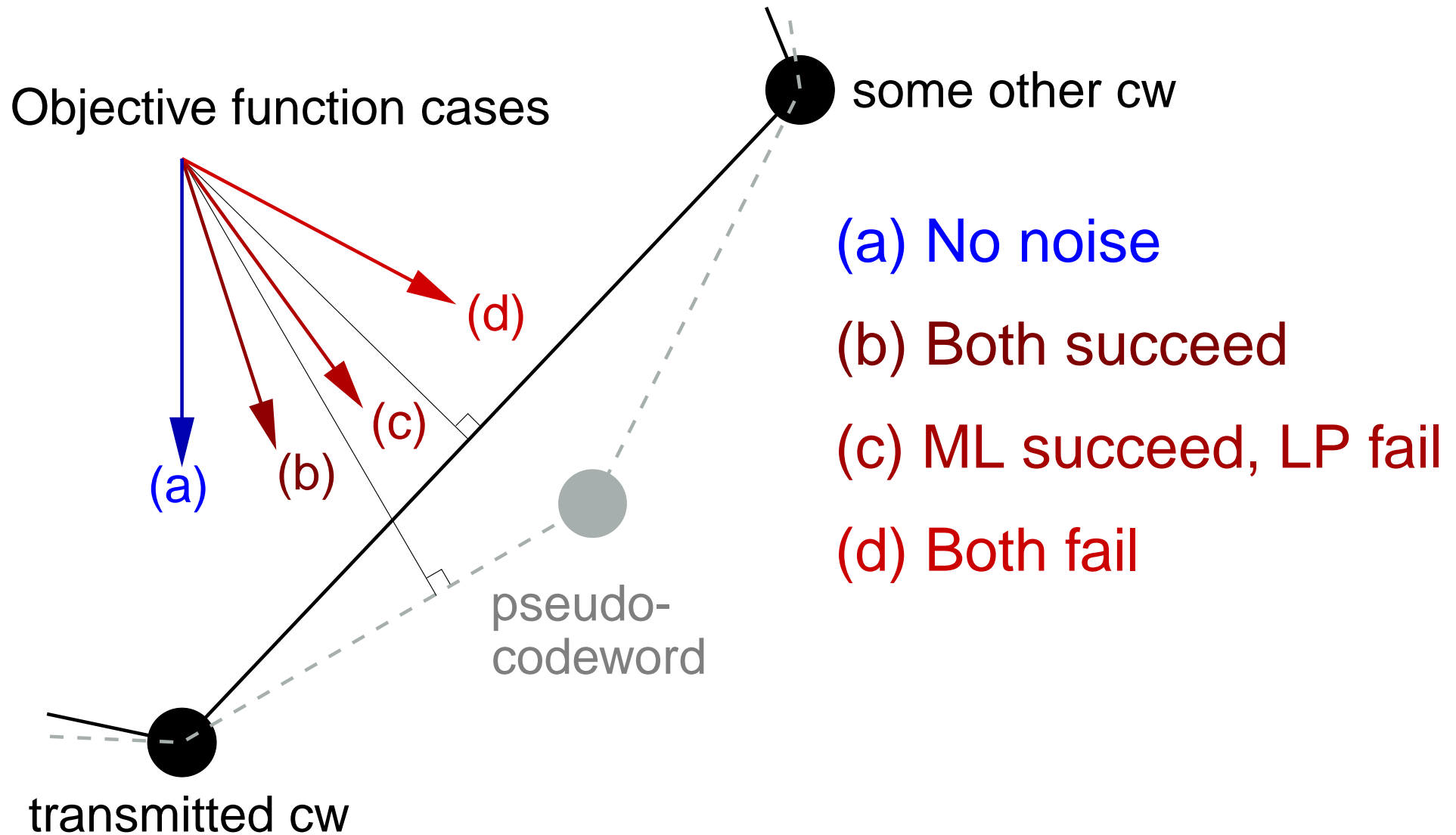
ML DECODING: **Given** corrupt codeword \hat{y} ,
find $y \in C$ such that $\sum_i \gamma_i y_i$ is minimized.

LP Decoding [FK '02, Fel '03]



- LP variables y_i for each code bit, relaxed $0 \leq y_i \leq 1$.
- Alg: Solve LP. If y^* integral, output y^* , else "error."
- *ML certificate* property

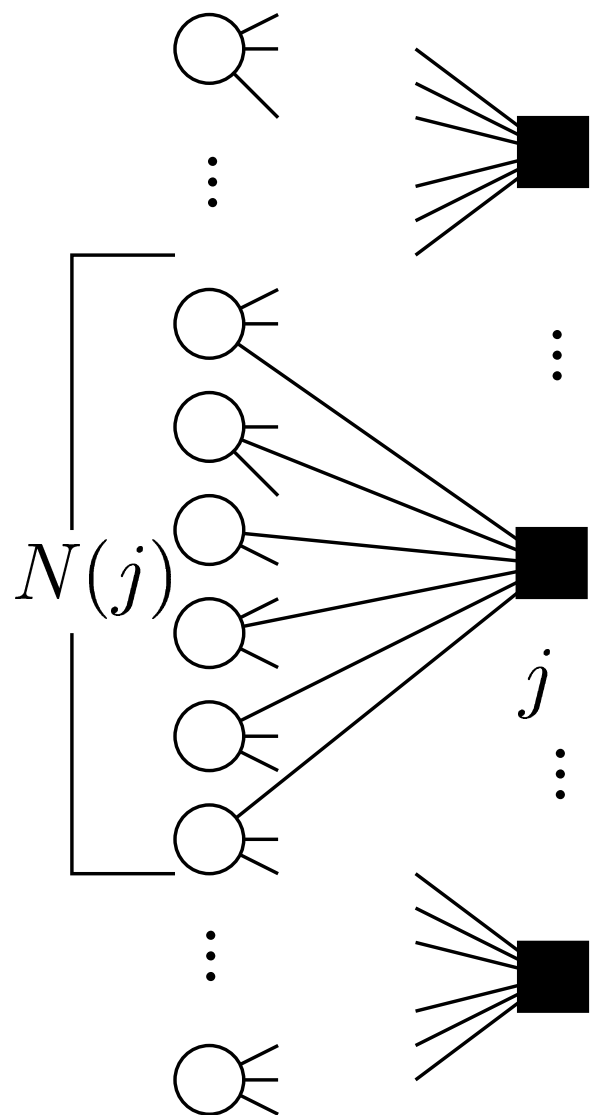
LP Decoding Success Conditions



Using a Dual Witness to Prove Success [FMSSW '04]

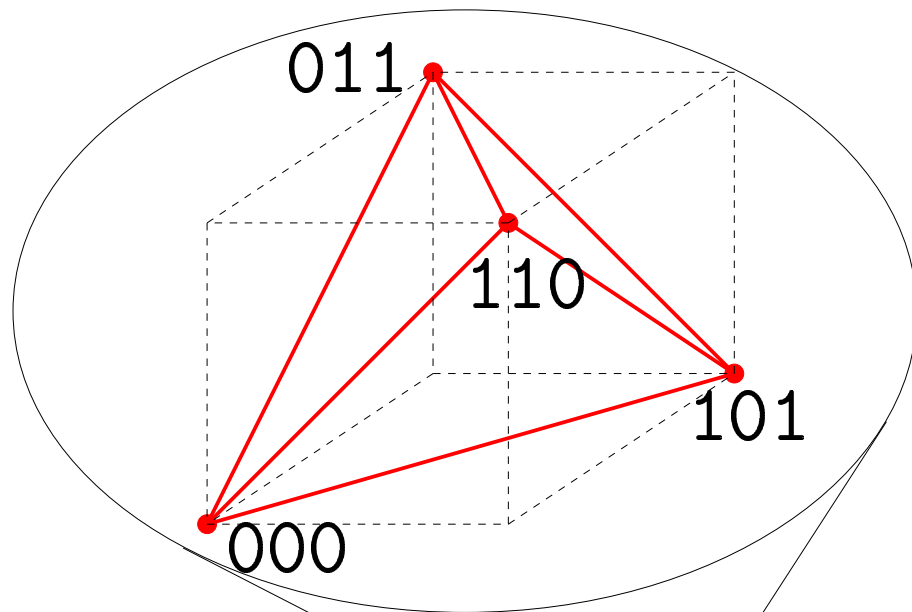
- Assume 0^n is transmitted (polytope symmetry); assume unique LP optimum (no problem).
 - success \iff Point 0^n is LP optimum
 - \iff \exists dual feasible point w/ value 0
- Take LP dual, set dual objective = 0: polytope \hat{P} .
 - success \iff \hat{P} non-empty
- Buys “analytical slack:”
 - ◆ With no noise \implies \hat{P} is large.
 - ◆ Noise increases \implies \hat{P} shrinks.
 - ◆ Trade off strength of result with ease of analysis.
 - ◆ Prove result for LDPC codes, adversarial channel [FMSSW '04]

Tanner Graph Codes



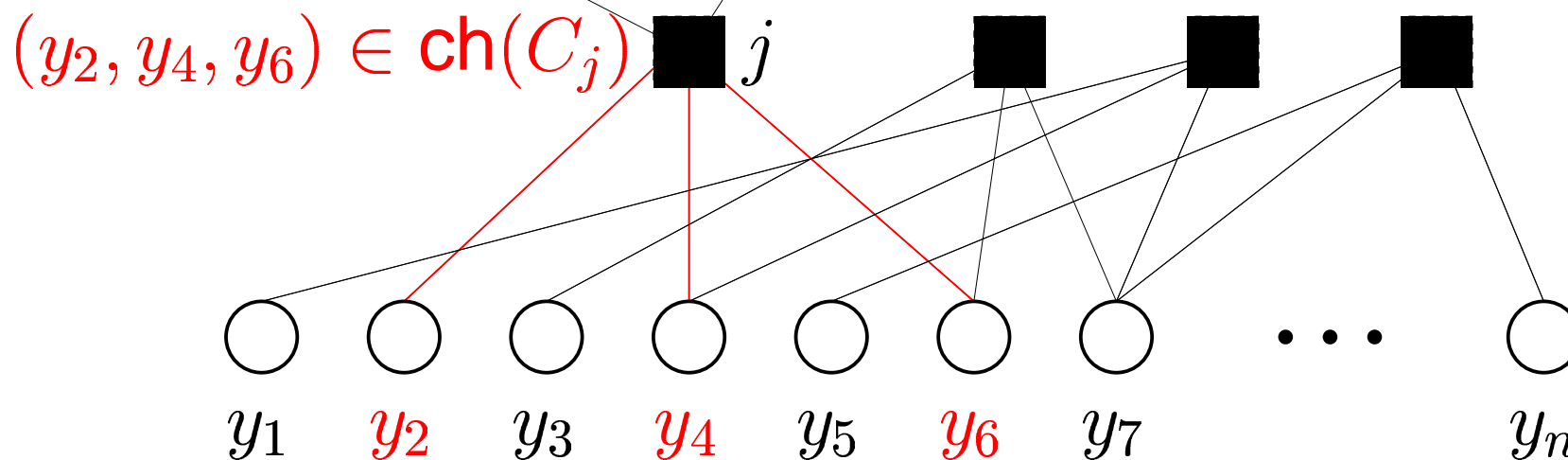
- Bipartite “Tanner” graph G :
 - ◆ n “variable” \bigcirc nodes
 - ◆ m “check” \blacksquare nodes
 - ◆ Subcode C_j for each check j .
 - ◆ Overall codeword: setting of bits to var nodes s.t.:
 - $\forall j$, bits of $N(j)$ in code C_j .
 - ◆ LDPC codes: special case w/ const. degree, $C_j =$ single parity check code.
- Ex: G is (3,6)-regular, $C_j = \{000000, 111000, 000111, 111111\}$.

LP Relaxation for Tanner Codes [FWK '03]

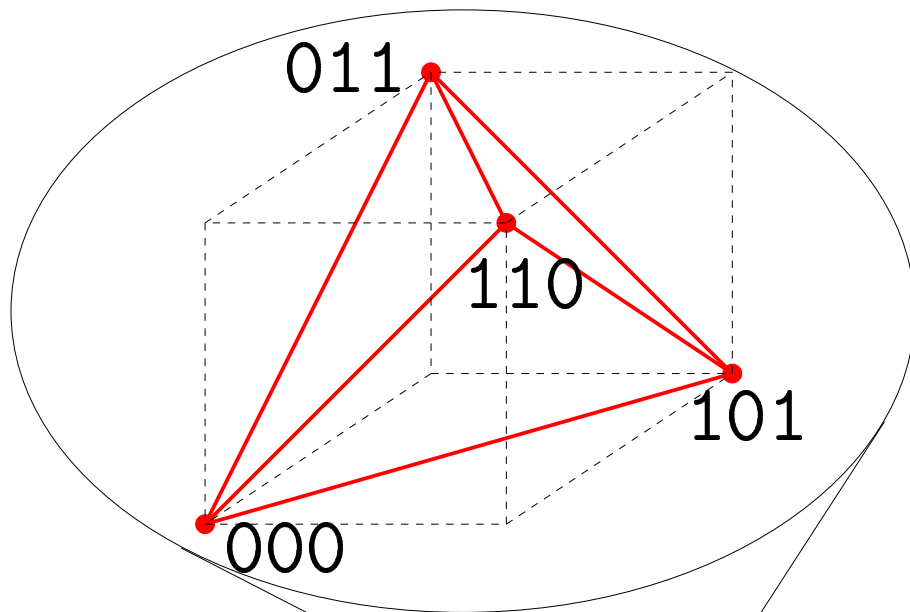


LP: $\min \sum_i \gamma_i y_i$ s.t.
For all check nodes j ,
 $\{y_i : i \in N(j)\} \in \text{ch}(C_j)$.

$\text{ch}(C_j)$ = convex hull
of local codewords.



LP Relaxation for Tanner Codes

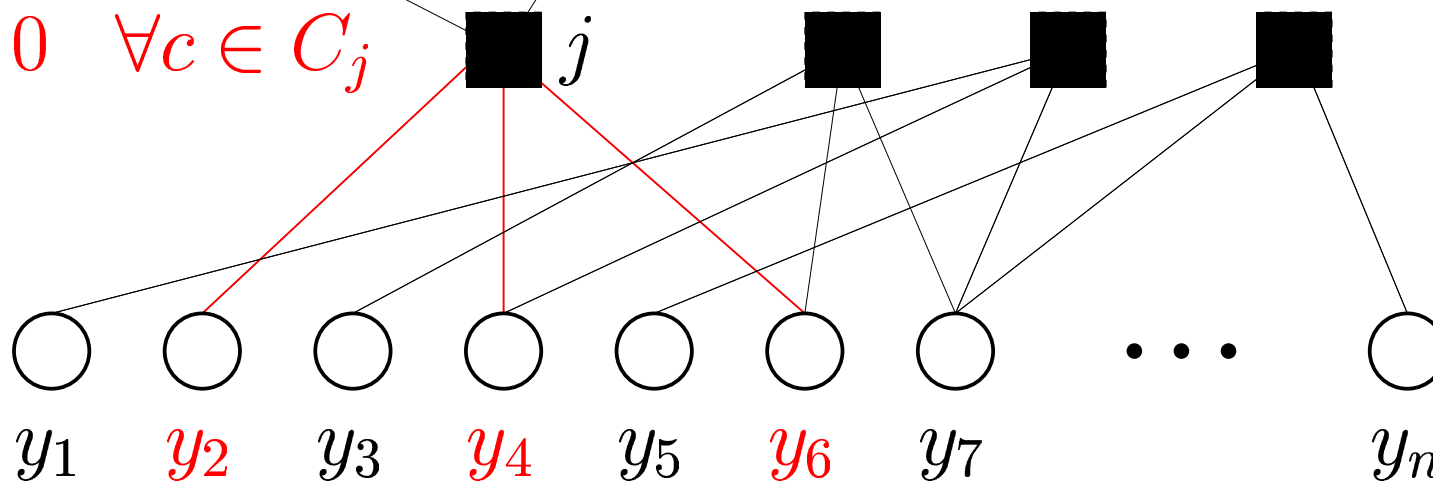


$$\text{LP: } \min \sum_i \gamma_i y_i \text{ s.t.}$$

$$\forall j, \quad \sum_{c \in C_j} w_{j,c} = 1$$

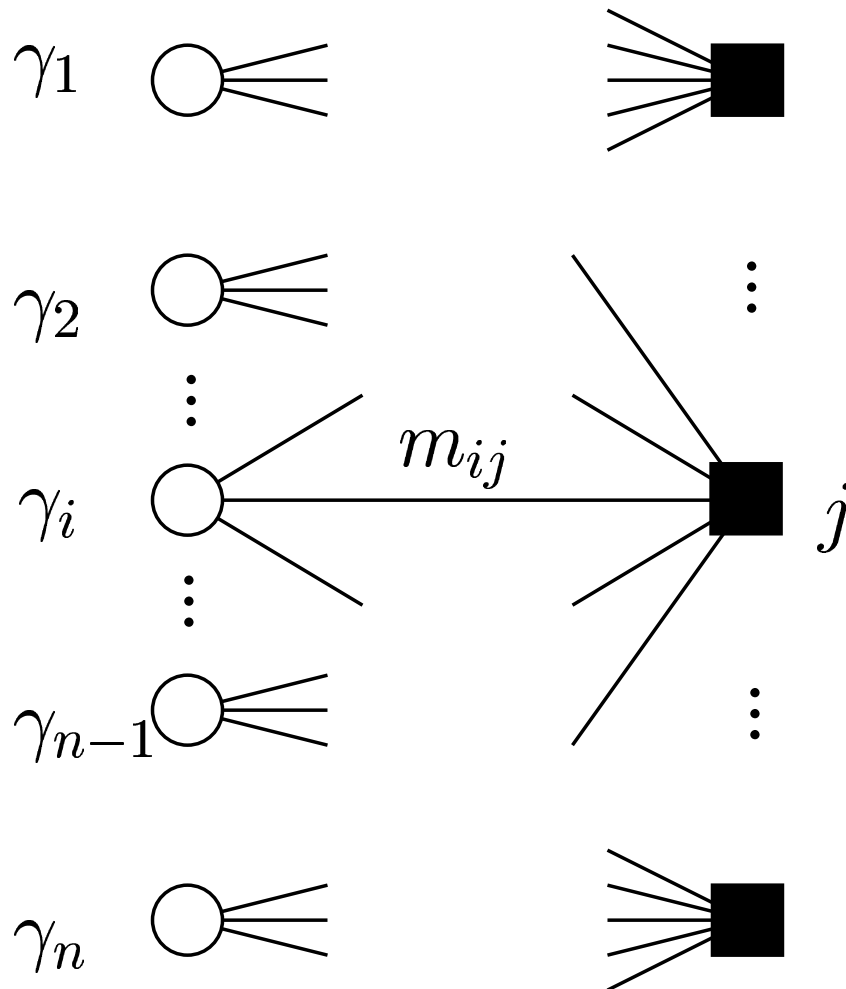
$$\forall (i, j), \quad y_i = \sum_{c: c_i=1} w_{j,c}$$

$$w_{j,c} \geq 0 \quad \forall c \in C_j$$



Dual Polytope: Tanner Codes

- Polytope \hat{P} for general Tanner codes:



- Edge weights m_{ij} .
- For all code bits (left nodes) i ,

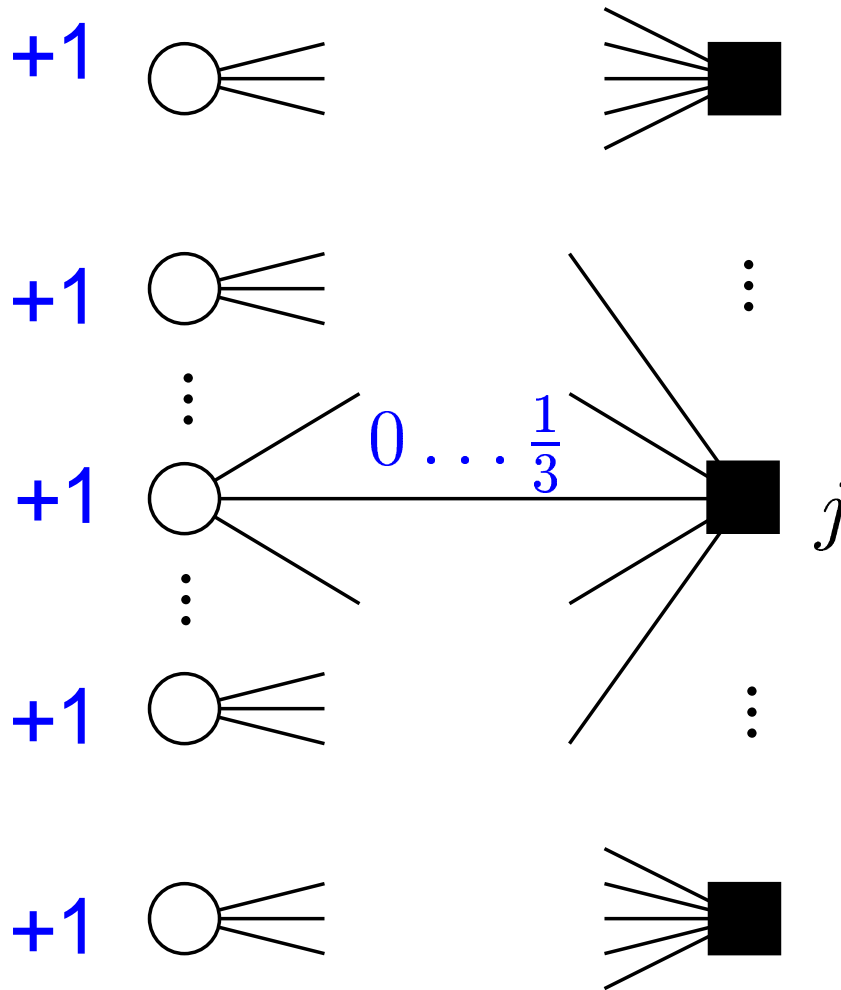
$$\sum_{j \in N(i)} m_{ij} \leq \gamma_i.$$

- For all checks j ,
codewords $c \in C_j$,

$$\sum_{i \in \text{sup}(c)} m_{ij} \geq 0$$

Dual Polytope: Tanner Codes

- No noise in the binary symmetric channel...



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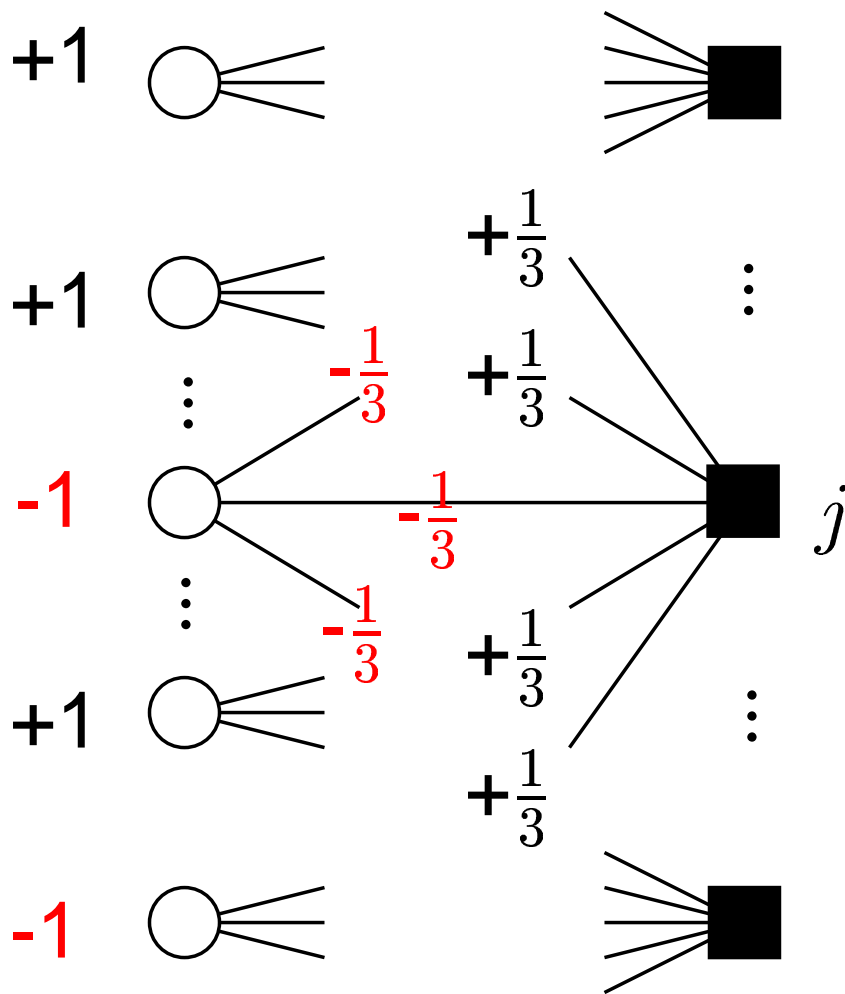
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Dual Polytope: Tanner Codes

A bit of noise...



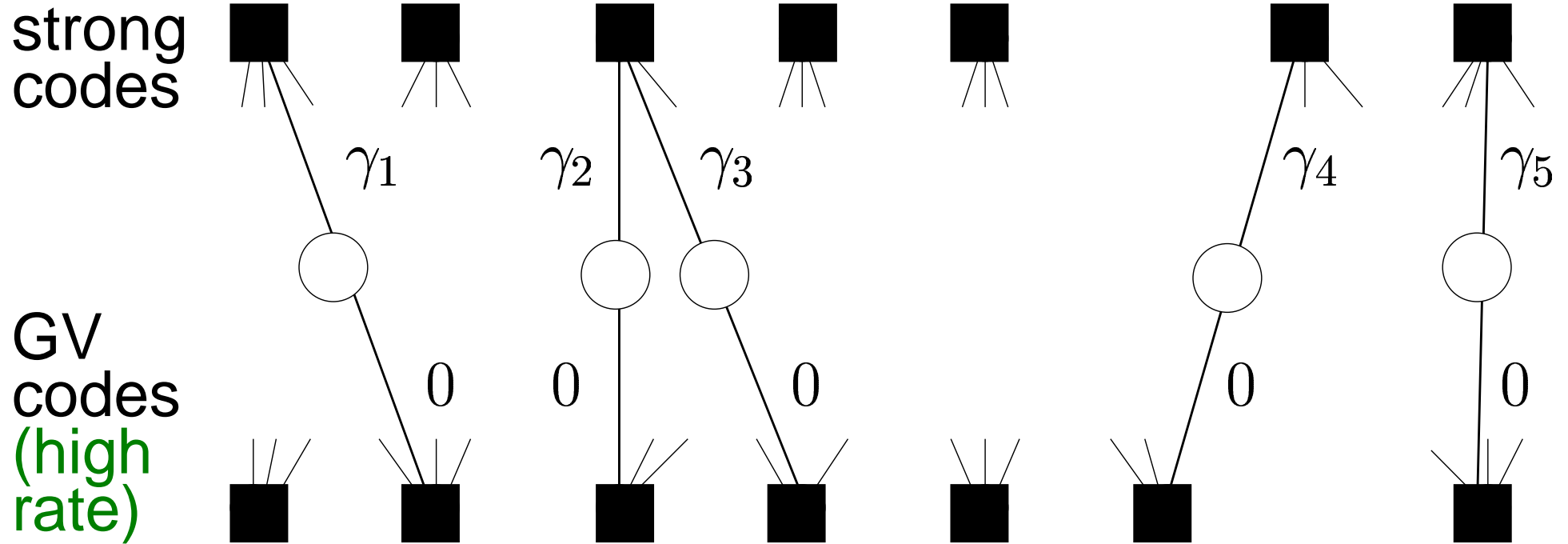
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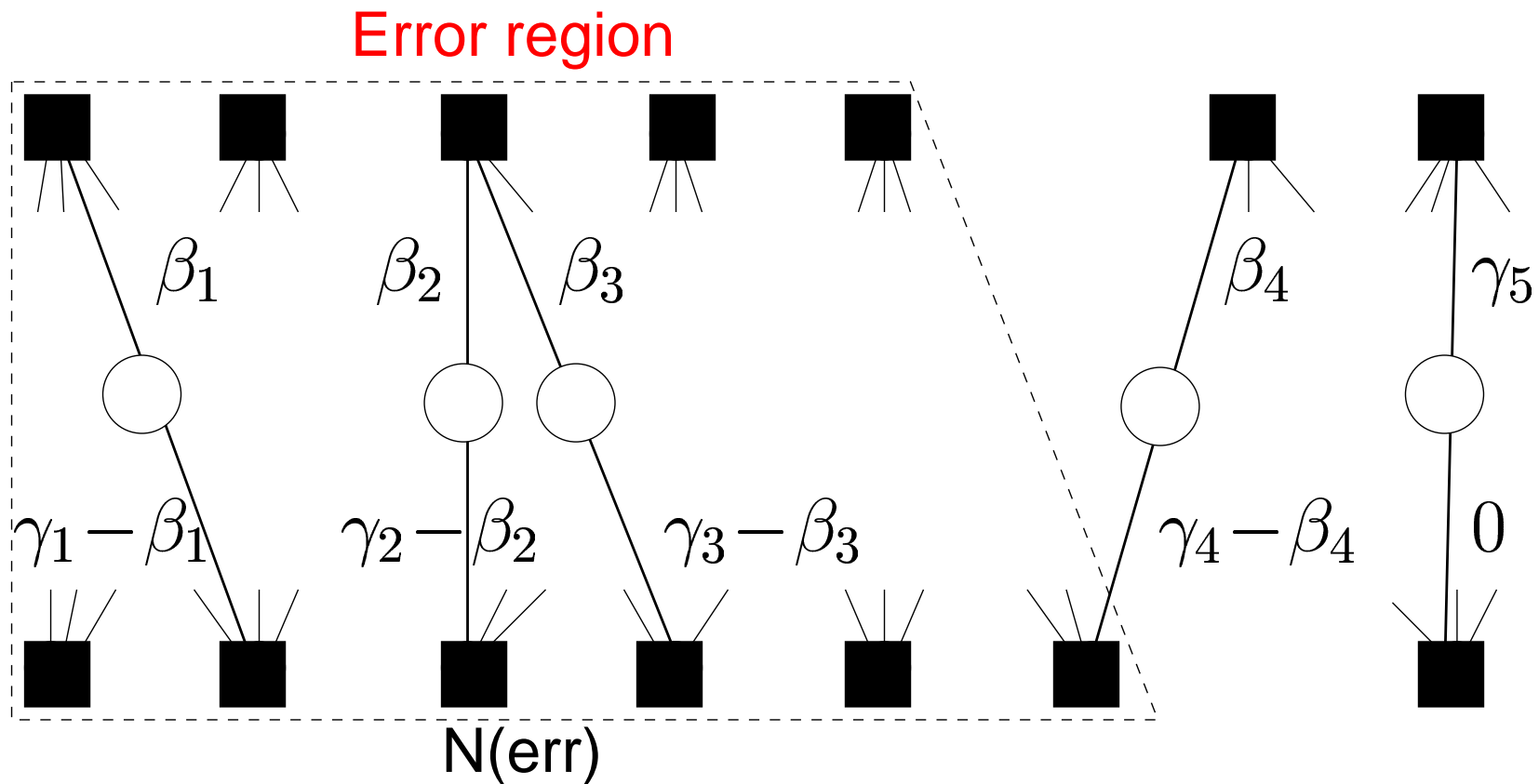
$$\sum_{i \in \text{sup}(c)} m_{ij} \geq 0$$

Using Expansion to Set Edge Weights



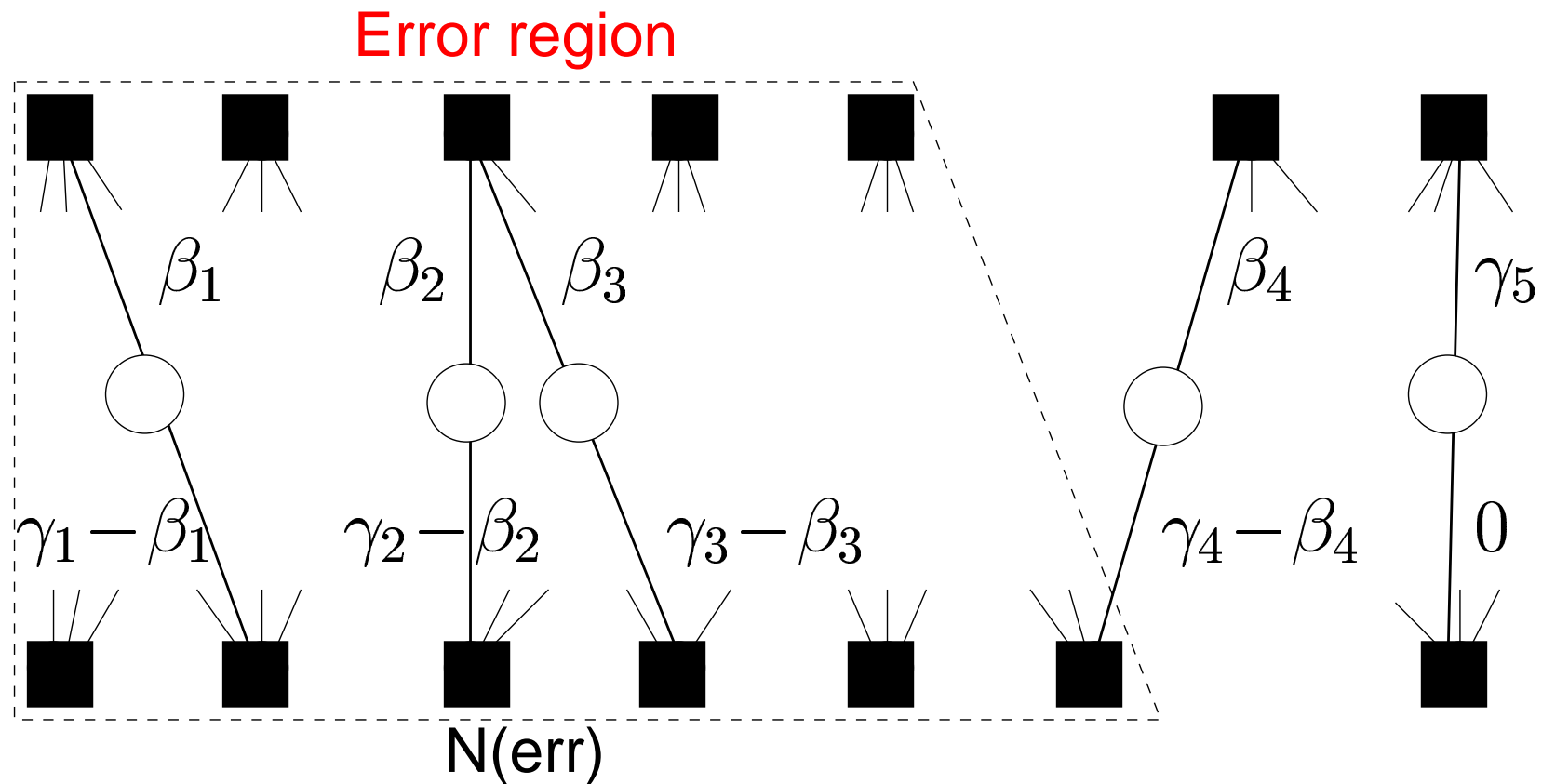
- Code built from Ramanujan graph [SS '96, BZ '02].
- Top nodes: strong codes, rate $r + \epsilon$
- Bottom nodes: GV-bound codes, rate $1 - \epsilon$
- **Initial weighting:** all “top” edges $m_{ij} = \gamma_i$, bottom edges $m_{ij} = 0$.

Using Expansion to Set Edge Weights



- “Error region:” checks w/ violated dual constraints.
- Use expansion to spread out excess weight among neighbors... details interesting but omitted.
- Maintain $m_{ij} + m_{ij'} = \gamma_i$, check constraints.

Using Expansion to Set Edge Weights



Thm: For any memoryless symmetric LLR-bounded channel with capacity \mathcal{C} , and **any rate** $r < \mathcal{C}$, there exists a code family of rate r for which the word error rate of LP decoding is $2^{-\Omega(n)}$.

Future work: LP Decoder Applications/Extensions

- Turbo codes: natural “flow-like” LP [FK '02][Fel '03]
 - ◆ RA(1/2) codes: $WER = n^{-\Omega(1)}$ [FK '02][EH '03]
 - ◆ Can we get $WER = 2^{-\Omega(n^\epsilon)}$? (e.g., rate 1/3 RA)

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-

- Tighter relaxation (lift and project)?
- ML decoding using IP (branch and bound)?
- Deeper connections to message-passing?
- More efficient algorithm to solve LP?

Major Open Questions

- Achieve capacity, decoding time $\text{poly}(n, \frac{1}{\mathcal{C}-r})$.
 - ◆ All previous capacity results:
Explicit ML decoding of $\text{poly}(\frac{1}{\mathcal{C}-r})$ -length codes.
 - ◆ This result:
Representation of $\text{ch}(\text{poly}(\frac{1}{\mathcal{C}-r})$ -length codes)

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- Achieve near-capacity rates for LDPC codes without the tree assumption.
 - ◆ “Small cycles are bad”: fact of fiction?
 - ◆ Expansion: to limited?