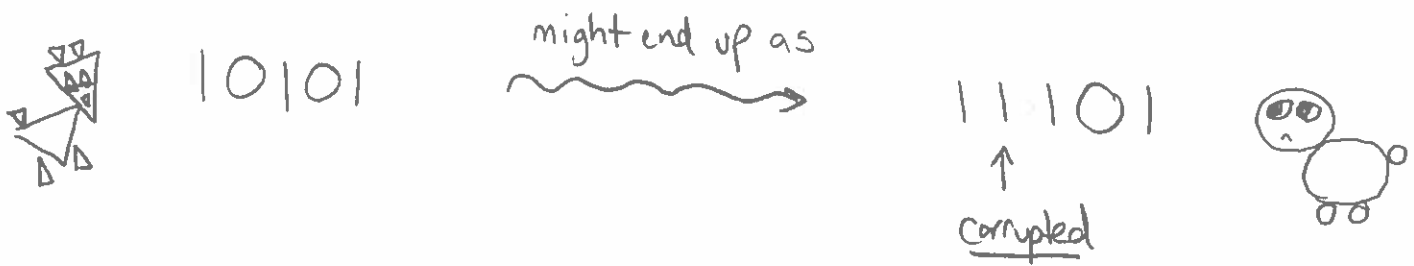


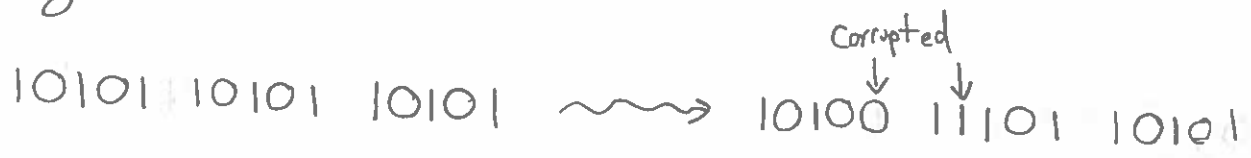
TURBO CODING

- ① Consider the task of communicating a message using a communication method that corrupts 5% of the bits you send, e.g.




How can you ensure that the recipient will understand what you sent?

- ② One approach is to send multiple redundant copies of the message:



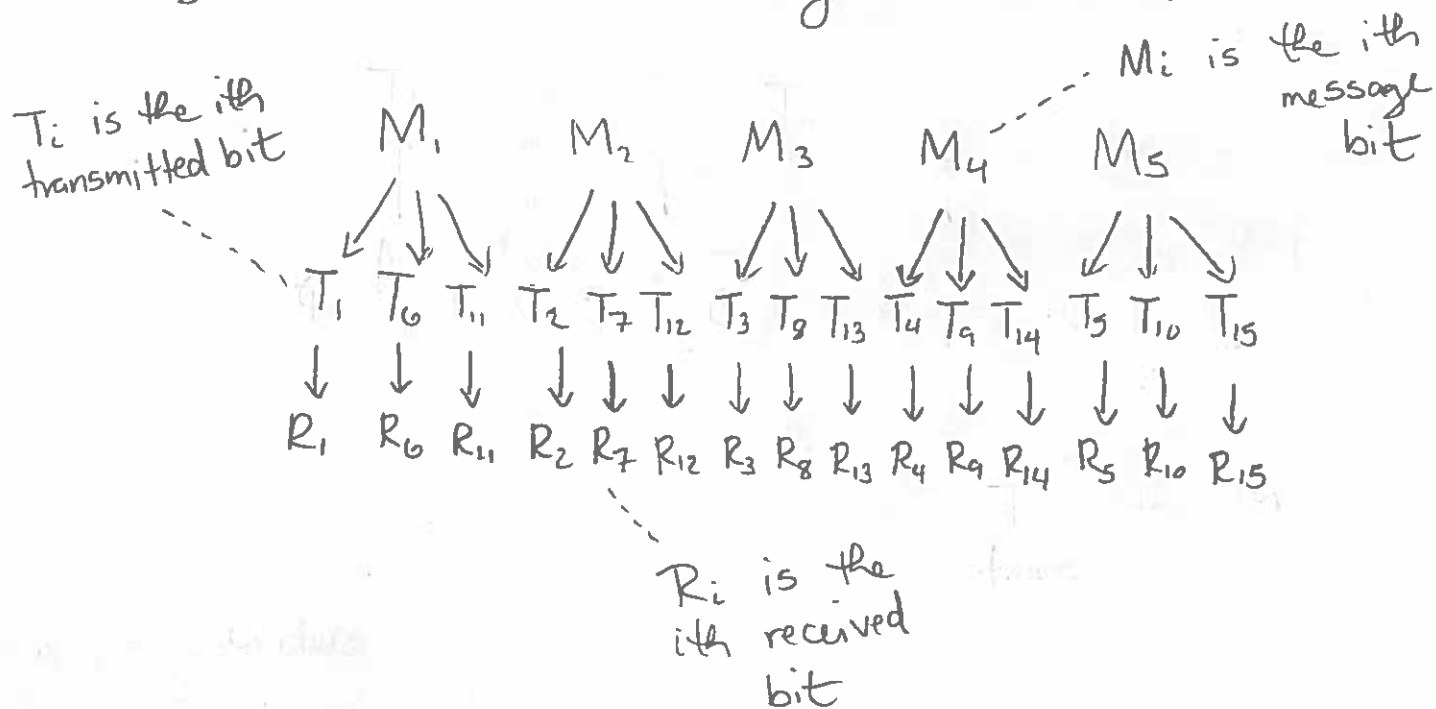
probably 10101
is the true message,
since it falls
"between" 10100 and
11101



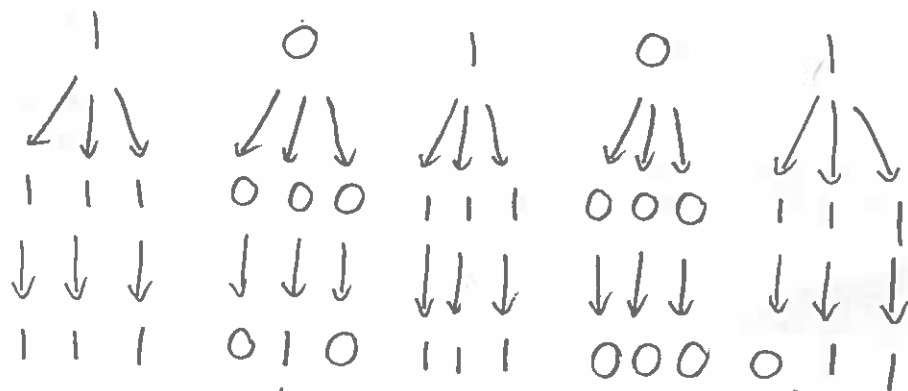
If you send enough redundant copies, then hopefully we can reconstruct the message.

TURBO CODING

③ How can we reconstruct the most probable original message? Model it as a Bayesian network!



e.g.



bit 7 is corrupted

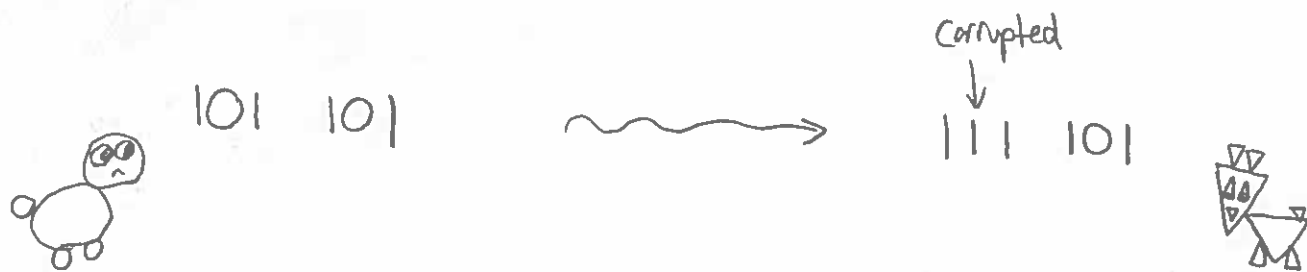
T_i	$P(R_i=1 t_i)$
0	.2
1	.8

bit 5 is corrupted

TURBO CODING

- ④ The recipient observes only the received bits R_1, \dots, R_{15} . To infer the most probable original message, the recipient computes $\operatorname{argmax}_{m_1, \dots, m_5} P(m_1, \dots, m_5 | r_1, \dots, r_{15})$ Using the Bayesian network.

- ⑤ But this may not be the most efficient way to encode the original message. Suppose we have a 3-bit message and our bandwidth only allows us to send 6 bits total. If we simply send two redundant copies of the message, what if a bit gets corrupted?



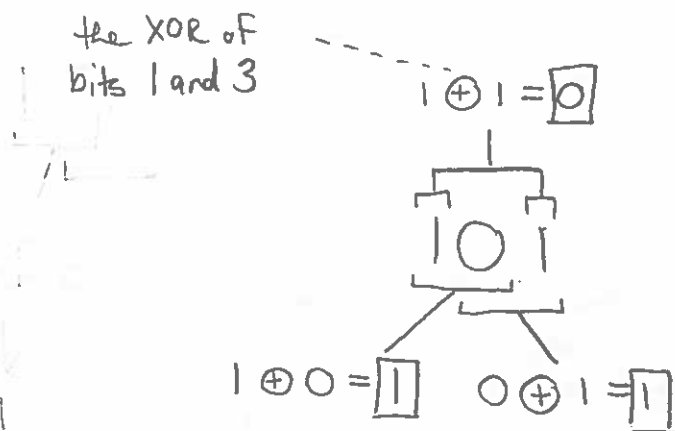
The received bits are equally likely given an alternative message:



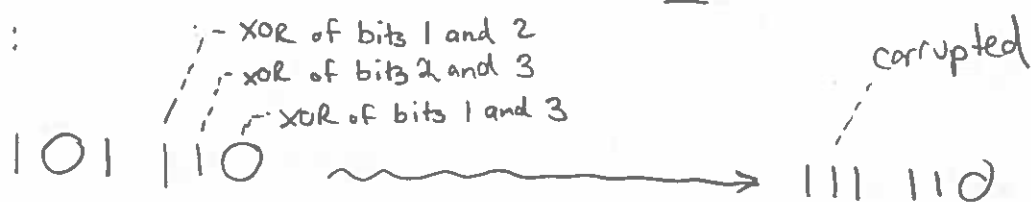
So if the recipient receives the transmission 111 101, there's no way to recover the original message.

TURBO CODING

⑥ What if we use our extra 3 bits in a more clever way?



Then:



But the message for 111 (with the same corruption) is:



Notice that the received bits 111 110 is closer to transmission 101 110 and received bits 101 000 is closer to transmission 111 000.

TURBO CODING

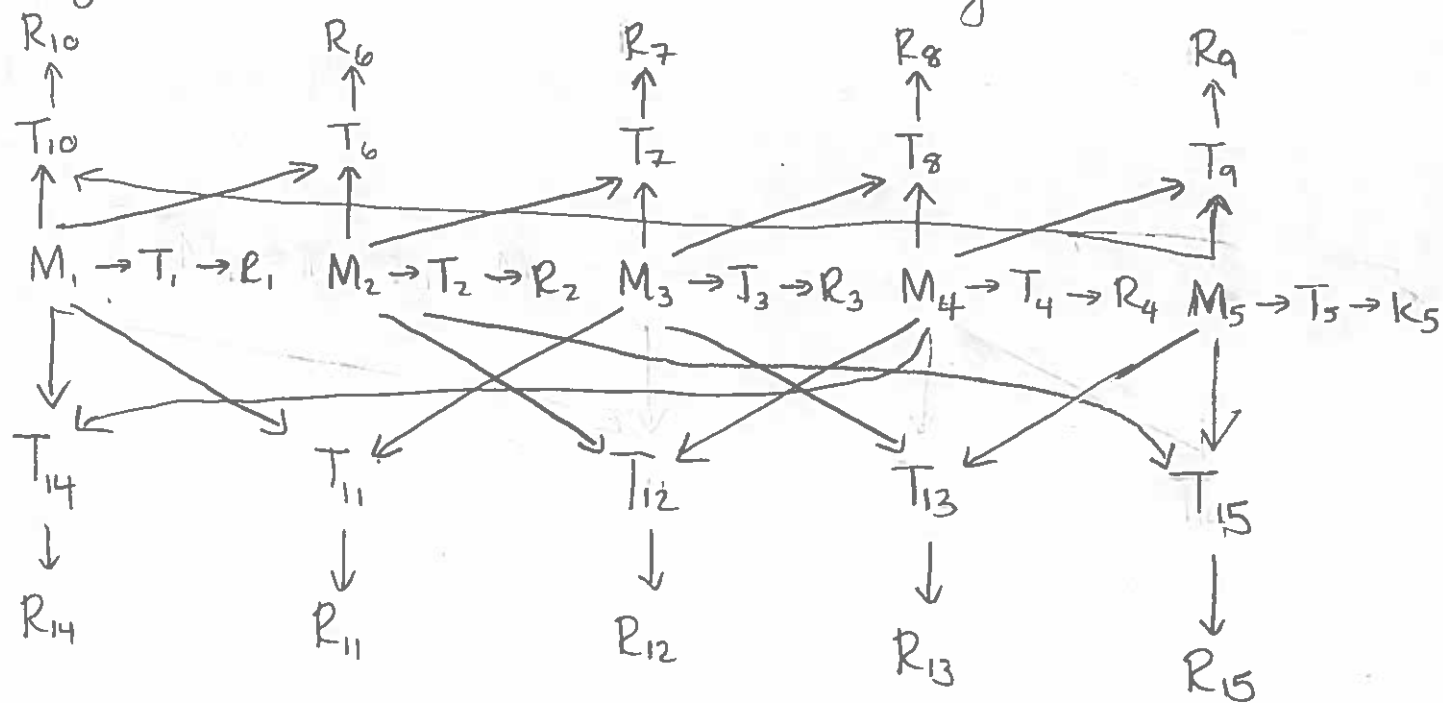
⑦ For an n -bit message, we can transmit as follows:

- let M_1, \dots, M_n be the original message bits
- let Transmission $T_i = M_i$ for $1 \leq i \leq n$.
- let transmission $T_{n+i} = M_i \oplus M_{(i+1) \% n}$ for $1 \leq i \leq n$
- let transmission $T_{2n+i} = M_i \oplus M_{(i+2) \% n}$ for $1 \leq i \leq n$

e.g. if the original message is 10100, then the transmission is:

10100 11101 01110

⑧ Again, we can model this as a Bayesian network:



And again, to infer the most probable original message, the recipient computes $\operatorname{argmax}_{m_1, \dots, m_5} P(m_1, \dots, m_5 | r_1, \dots, r_{15})$.

TURBO CODING

- ⑨ If we code up these Bayesian networks (let's call them RedundantNet and ParityNet), then we see the following statistics:

If the corruption probability is 0.10%:

	$P(\text{orig message} \text{received bits})$	
	<u>in RedundantNet</u>	<u>in ParityNet</u>
no corrupted bits	.993	.999
1 corrupted bit	.895	.997
2 corrupted bits	.807	.981

- ⑩ We can make ParityNet arbitrarily reliable by increasing the number of parity bits transmitted.

A very popular architecture that uses these ideas for encoding and decoding is called "Turbo Coding", which is a core technology behind 3G and 4G communications standards.

A seminal paper by Robert McEliece et al, called "Turbo Decoding as an Instance of Pearl's Belief Propagation Algorithm" describes the connections between turbo coding and Bayesian network inference.