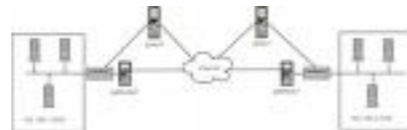


# Coding Theory: Achievements and Challenges



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Noisy Channel →



# Channel Model (Simplified)

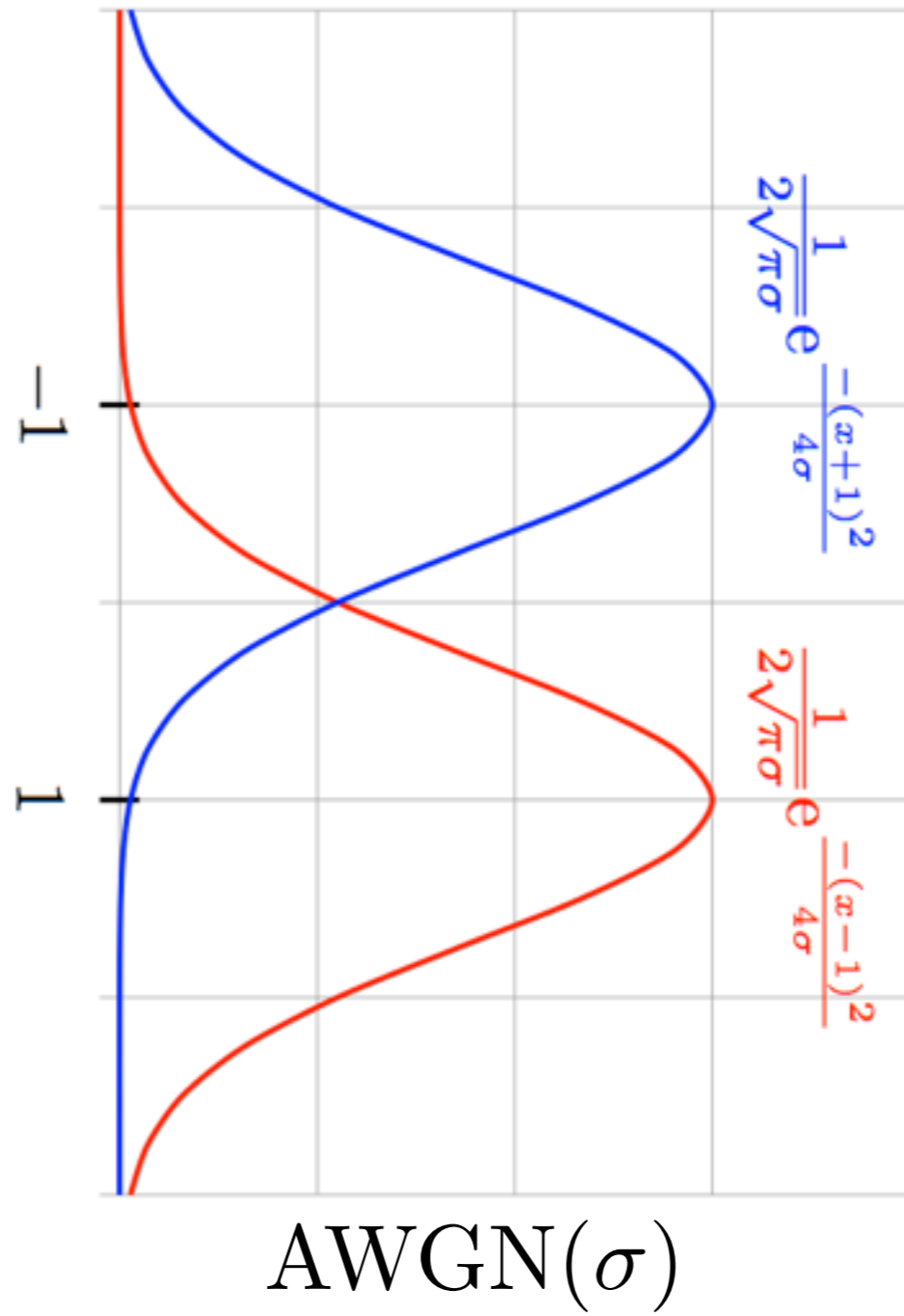
$\Sigma, \Gamma$  finite alphabets

$p: \Sigma \times \Gamma \rightarrow \mathbb{R}$  Conditional probability distribution

$p(x | y)$  Probability that  $x$  is sent given that  $y$  is received

For fixed  $x$   $p(x | y)$  is a probability distribution on  $\Sigma$

# Examples ( $\Sigma = \text{GF}(2)$ )



# Entropy, Mutual Information, Capacity

$X$  r.v. on  $\Sigma$ ,  $Y$  r.v. on  $\Gamma$

$$H(X) = - \sum_{x \in X} \Pr[X = x] \log_2(\Pr[X = x])$$

Entropy

How many “bits of uncertainty” does  $X$  have?

# Entropy, Mutual Information, Capacity

$X$  r.v. on  $\Sigma$ ,  $Y$  r.v. on  $\Gamma$

$$H(X) = - \sum_{x \in X} \Pr[X = x] \log_2(\Pr[X = x])$$

$$H(X|Y) = - \sum_{x \in \Sigma, y \in \Gamma} \Pr[X = x, Y = y] \log_2(\Pr[X = x|Y = y])$$

**Conditional Entropy**

How many “bits of uncertainty” does  $X$  have, if we know  $Y$ ?

# Entropy, Mutual Information, Capacity

$X$  r.v. on  $\Sigma$ ,  $Y$  r.v. on  $\Gamma$

$$H(X) = - \sum_{x \in X} \Pr[X = x] \log_2(\Pr[X = x])$$

$$H(X|Y) = - \sum_{x \in \Sigma, y \in \Gamma} \Pr[X = x, Y = y] \log_2(\Pr[X = x|Y = y])$$

$$I(X; Y) := H(X) - H(X|Y)$$

**Mutual Information**

What is the reduction of “bits of uncertainty” of  $X$  if we know  $Y$ ?

# Entropy, Mutual Information, Capacity

$X$  r.v. on  $\Sigma$ ,  $\mathcal{C}$  channel with law  $p(x | y)$

$Y$  r.v. induced on  $\Gamma$  by  $X$ .

$$q(x) = \Pr[X = x]$$

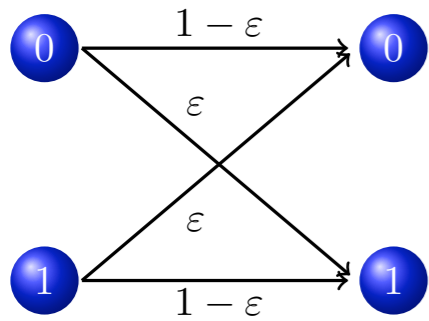
$$\text{Cap}(\mathcal{C}) = \max_q I(X; Y)$$

**Channel Capacity**

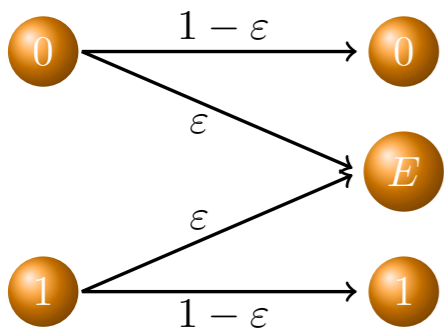
Best “reduction of uncertainty” of  $X$  given  $Y$



## Examples



$$\begin{aligned} \text{Cap}(\text{BSC}(\varepsilon)) &= 1 + \varepsilon \log_2(\varepsilon) + (1 - \varepsilon) \log_2(1 - \varepsilon) \\ &= 1 - h(\varepsilon) \end{aligned}$$



$$\text{Cap}(\text{BEC}(\varepsilon)) = 1 - \varepsilon$$

# Shannon's Channel Coding Theorem

Reliable communication over channel  $\mathcal{C}$  is possible for any rate  $R < \text{Cap}(\mathcal{C})$

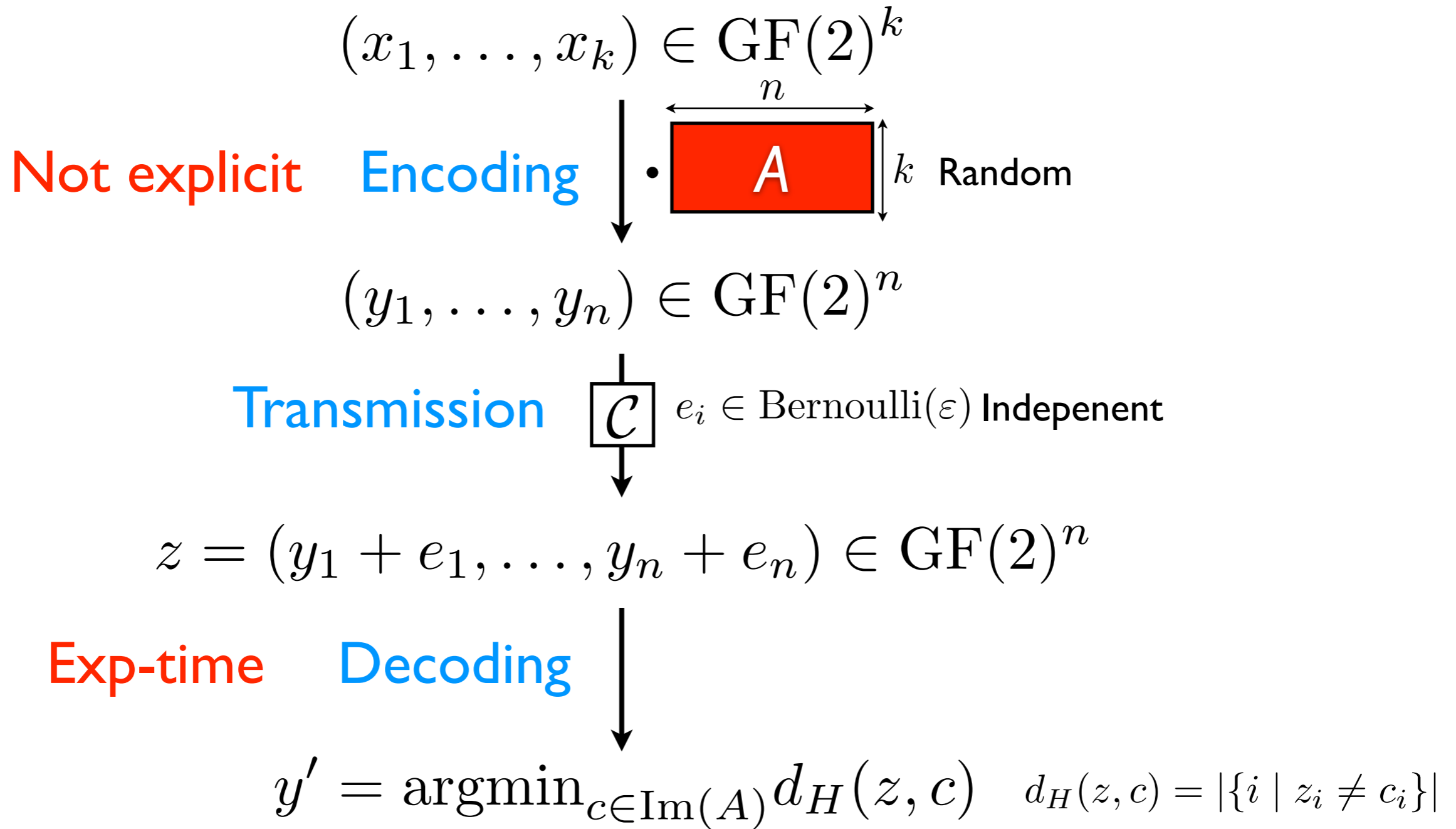
↑  
Average number of bits sent per channel use

Impossible if  $R > \text{Cap}(\mathcal{C})$



Claude E. Shannon  
1916-2001

# Example: BSC



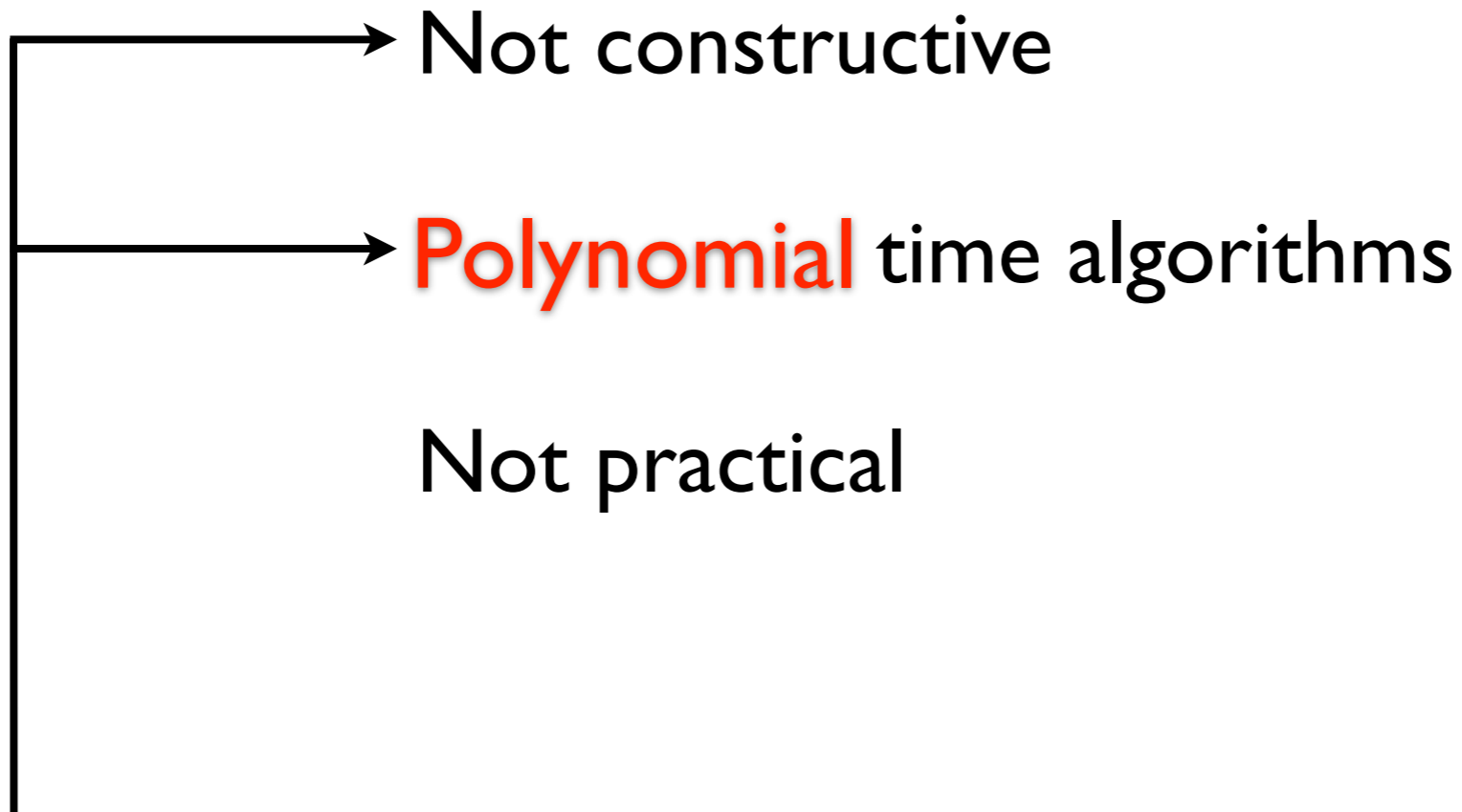
# Example: BSC

If  $A$  is chosen randomly, and  $k/n < 1 - h(\varepsilon)$  then

$$\frac{1}{2^k} \sum_{x \in \text{GF}(2)^k} \Pr[y' \neq x \cdot A] \leq \exp(-\gamma n)$$

Positive, depends on  $\varepsilon$  and  $k/n$

# Shannon's Channel Coding Theorem



Forney, 1967: **Concatenated Codes**

# Open Problem

For any given channel, design “practical” codes that come arbitrarily close to the capacity of that channel

# Linear Codes

# Bounds



# RS-Codes

# WB-Decoder

# List-Decoding

# AG-Codes

# LDPC Codes

# Decoding on the BEC

# Theory

# Theory



# Theory

# Achieving Capacity

# Other Channels

# Other Channels

# Other Channels

# Further Developments