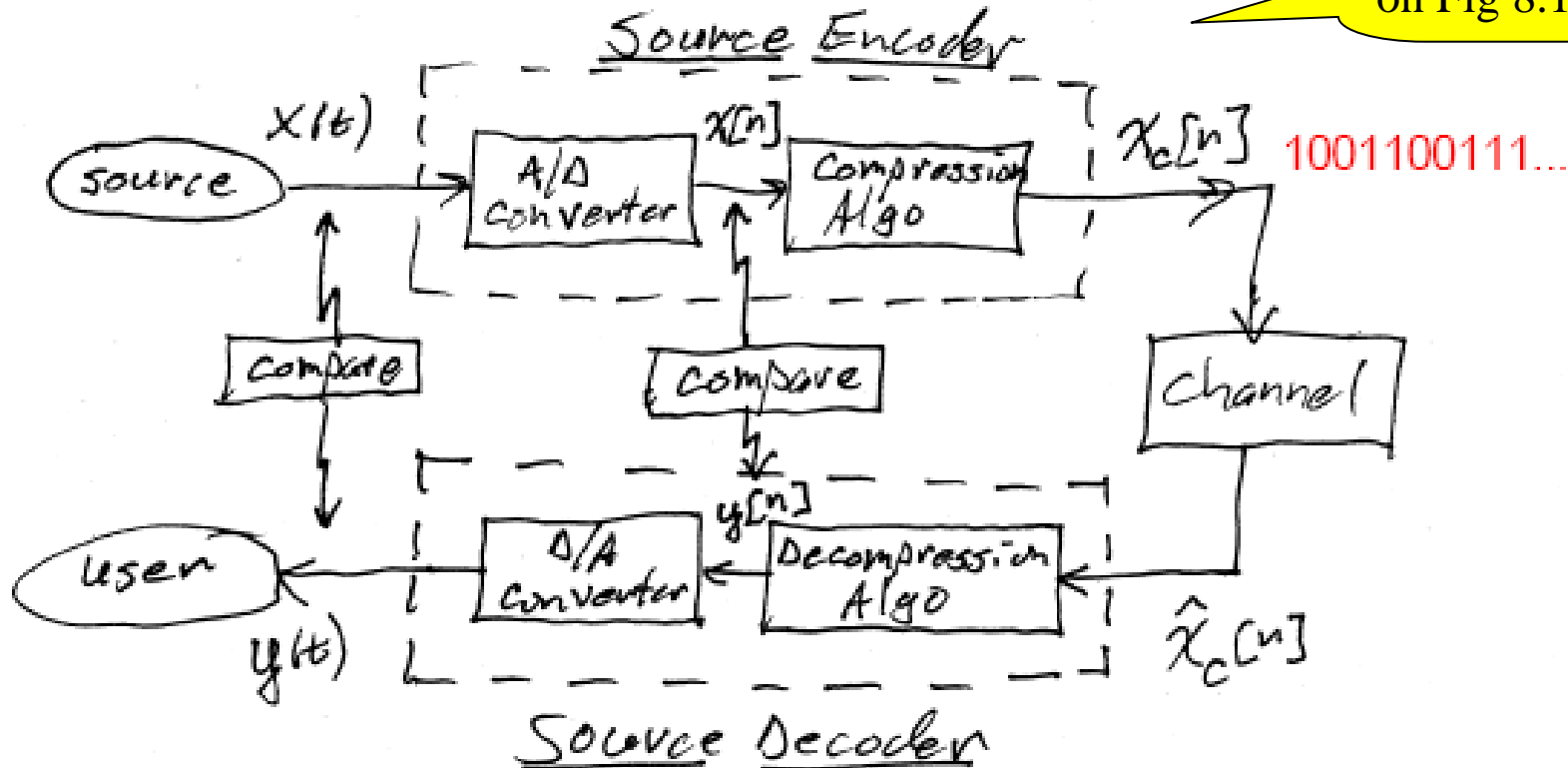


# Ch. 8 Math Preliminaries for Lossy Coding

## 8.3 Distortion Criteria or Measure

# Structure of Lossy Coding

Recall Slight variation on Fig 8.1 in textbook:



In practice this is how things get implemented. The A/D not only samples  $x(t)$  in time but it also discretizes the values – however, the discretization is very fine. The Comp. Algo. often includes further, coarser discretization

In theory, we think of the A/D as only sampling the signal in time – the Comp. Algo. handles the discretization. Thus, we often think of:

$x[n]$  as taking values on continuum

$y[n]$  as taking discrete values

# Comparing Original & Compressed Signals

- How do we check how close  $y[n]$  is to the original  $x[n]$ ?
- We must define a distortion measure  $d(x,y)$

Most Common:  $d(x, y) = (x - y)^2$

Square Error  
(SE) Measure

- Now, usually we have  $N$  samples to compare, so we use:

$$d(\mathbf{x}, \mathbf{y}) = \frac{1}{N} \sum_{n=1}^N d(x[n], y[n])$$

Vectors of  
Samples

An “operational”  
Mean Square Error  
(MSE) Distortion

$$= \frac{1}{N} \sum_{n=1}^N (x[n] - y[n])^2$$

If SE measure  
is used

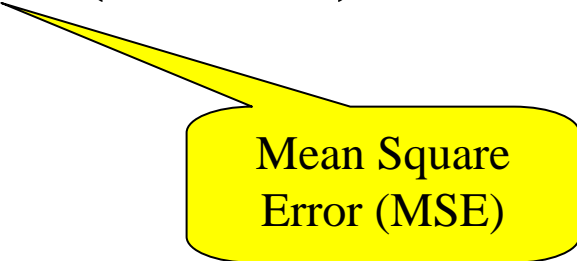
- In practice we’ll want to adapt comp. algo. to give the smallest value of  $d(\mathbf{x}, \mathbf{y})$  for the particular  $\mathbf{x}$  you are processing...

“Operational Distortion” Viewpoint

- In Theory we don't have a particular  $\mathbf{x}$  (in a theoretical setting we haven't collected a signal yet) so we strive to minimize  $d(\mathbf{x}, \mathbf{y})$  on average... a probabilistic average over the ensemble of  $\mathbf{x}$ 's according to some probability model.

$$\begin{aligned} D &= E \{d(\mathbf{x}, \mathbf{y})\} \\ &= \frac{1}{N} \sum_{n=1}^N E \{d(x[n], y[n])\} \\ &= E \{d(x, y)\} \quad \text{if stationary process} \end{aligned}$$

If SE is used (& stationary):  $D = E \left\{ (y - x)^2 \right\} = \sigma_{err}^2$



Mean Square  
Error (MSE)

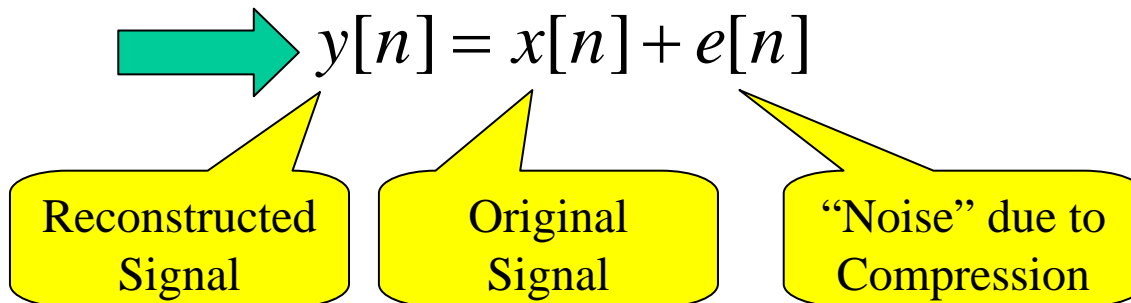
# Non-MSE Distortion Measures

- MSE is the “most widely” used due to its simplicity of application (math results are fairly easy to derive)
- But SE doesn’t always correspond well with visual/audio quality as perceived by humans
  - Compression algorithms intended for video/audio often use distortion measures that include ways to capture the psychology of human vision/hearing
- MSE is usually not the best choice when the decompressed signal is going to be used in statistical estimation/decision processing
  - See many of my papers posted on my web page
- **So why study MSE?**
  - Math is easy (relative to that needed for non-MSE)
  - It gives decent results
  - It is usually part of the non-MSE measures

# Relating Distortion To SNR

SNR = Signal-to-Noise (Power) Ratio

Since...  $d(x[n], y[n]) = \underbrace{(y[n] - x[n])^2}_{e[n]}$



The reconstructed signal has

an SNR of

$$\begin{aligned} SNR &= \frac{\text{power}\{x[n]\}}{\text{power}\{e[n]\}} \\ &= \frac{\sigma_x^2}{\sigma_e^2} = \frac{\sigma_x^2}{D} \end{aligned}$$



$$SNR = \frac{\sigma_x^2}{D}$$

**SNR has the advantage that it measures distortion relative to the signal power**

SNR is typically stated in dB form:

$$SNR_{dB} = 10 \log_{10} \frac{\sigma_x^2}{\sigma_e^2}$$

In image compression it is common to use peak SNR:

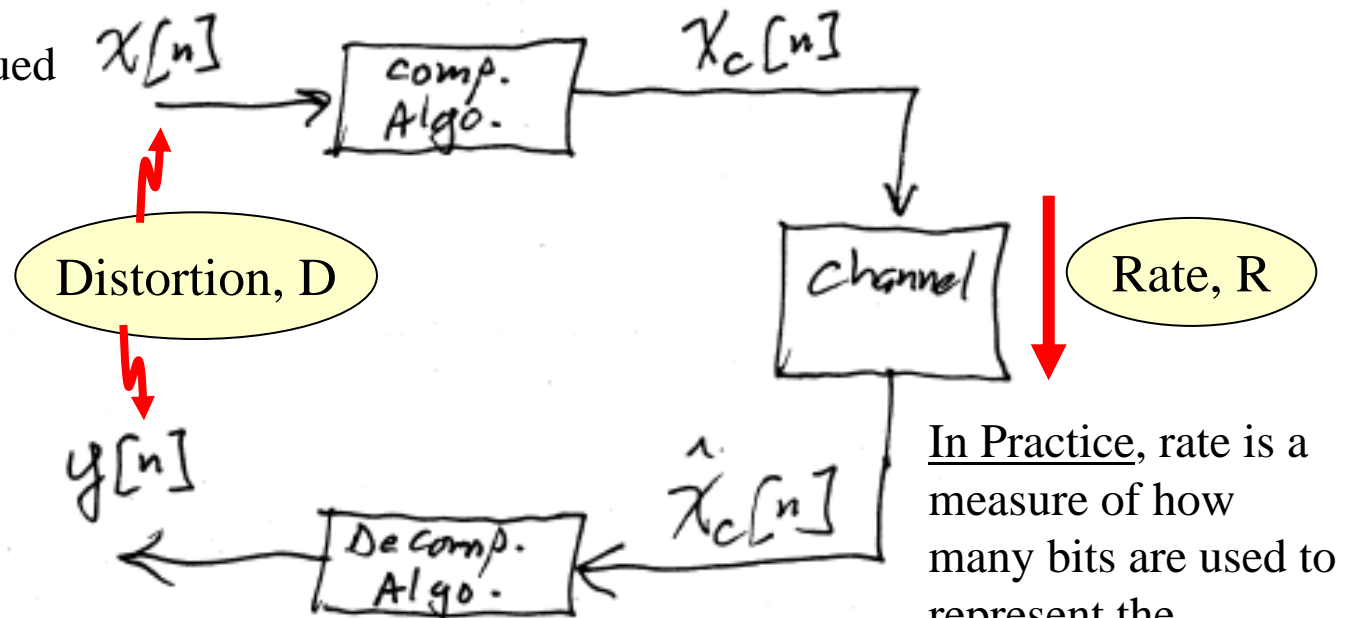
$$PSNR_{dB} = 10 \log_{10} \frac{x_{peak}^2}{\sigma_e^2}$$

$$x_{peak} = \max_n |x[n]|$$

# Goals of Lossy Compression

Think of  $x[n]$  as DT and Continuum-Valued (in practice,  $x[n]$  is “finely” discrete-valued)

Think of  $y[n]$  as DT and Discrete-Valued



In Practice, rate is a measure of how many bits are used to represent the compressed signal:

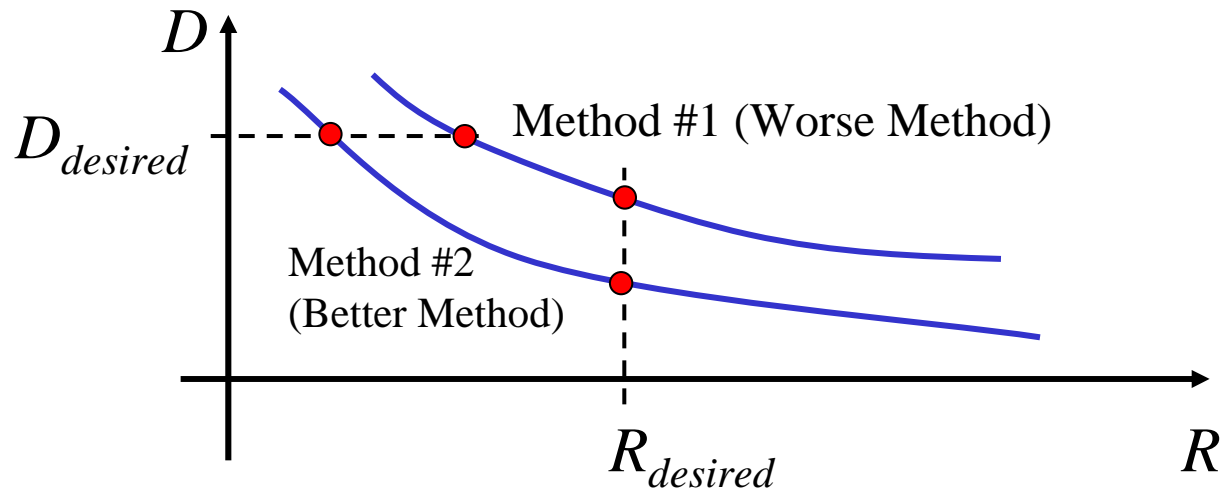
- bits/sample
- bits/second
- bits/fixed duration
- Etc.

## Goals of Compression

1. Reduce distortion for a fixed rate
2. Reduce rate for a fixed distortion

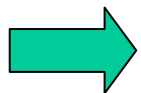


# Goals of Lossy Compression



## Goals of Compression

1. For a given  $R_{desired}$ , find a method that minimizes  $D$
2. For a given  $D_{desired}$ , find a method that minimizes  $R$



**Designing Lossy Comp. Algorithms involves solving constrained optimization problems**