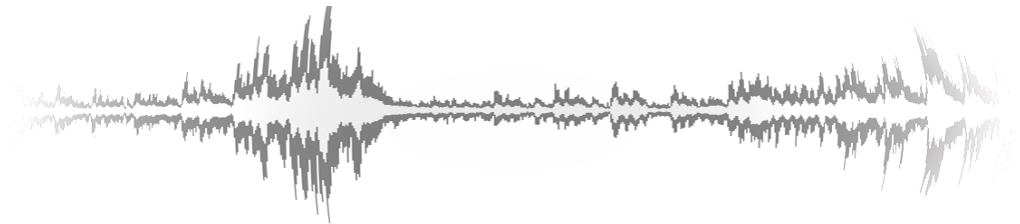


# Digital Signal Processing for Music

## Part 23: Source Coding

alexander lerch



# source coding

## introduction 1/3

- typical audio **bit rates**

$$16 \text{ bit} \cdot 44100 \text{ sps} \cdot 2 \text{ chan} = 1411.2 \text{ kbps}$$

$$24 \text{ bit} \cdot 192000 \text{ sps} \cdot 5 \text{ chan} = 23040 \text{ kbps}$$

- reasons for bit rate reduction

- economical reasons: cheaper transmission/storage
- technical reasons: restricted storage/transmission bandwidth

- applications for source coding

- Internet: streaming, distribution, peer-2-peer, VoIP, ...
- Media: DVD-V/A, ...
- Portable Devices: MP3-Player, cell phones, Mini-Disc, ...
- Broadcasting: (Digital) Radio, TV, ...
- Cinema: DD, DTS, SDDS
- ...

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# source coding

## introduction 2/3

**How can the bitrate be reduced**



# source coding

## introduction 2/3

### How can the bitrate be reduced



- ④ **lossless:**
  - remove *redundant* information (unnecessary to reconstruct the signal)
    - entropy coding
    - (linear predictive coding)

# source coding

## introduction 2/3

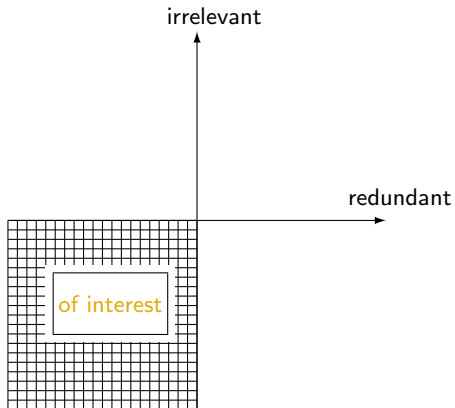
## How can the bitrate be reduced



- ① **lossless:**  
remove *redundant* information (unnecessary to reconstruct the signal)
  - entropy coding
  - (linear predictive coding)
  
- ② **lossy:**  
remove *irrelevant* information (not “missed” by the recipient)
  - waveform coding
  - perceptual coding

# source coding

## introduction 3/3





# source coding

## fundamentals: definitions

note: words to be transmitted are referred to as *symbols*

### information content

The less frequent a symbol, the higher its *information content, self-information, surprisal*.

$$I_n = \log_2 \left( \frac{1}{p_n} \right)$$

### entropy

The entropy is the *Expected Value* of the information content. It is the *theoretic minimum of bits* required for transmission.

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## fundamentals: information content and entropy examples

- **dice:**  $p_n = \frac{1}{6}$

$$I_n = \log_2 \left( \frac{1}{p_n} \right) = 2.58 \text{ bit}$$

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- **imperfect dice:**  $p_0 = \frac{1}{2}, p_{1\dots5} = \frac{1}{10}$

$$I_1 = \log_2(2) = 1 \text{ bit}$$

$$I_{2\dots6} = \log_2(10) = 3.32 \text{ bit}$$

$$H = \frac{1}{2} \cdot 1 + \frac{5}{10} \cdot 3.32 = 2.16 \text{ bit}$$

# source coding

## entropy coding: example 1

### idea: use shorter words for frequent symbols

- 3 possible symbols

symbol	probability	word
A	$p = 0.5$	
B	$p = 0.25$	
C	$p = 0.25$	

- entropy

$$H = \sum_{n=0}^{N-1} p_n \log_2 \left( \frac{1}{p_n} \right) = 1.5$$

- transmit the following group of symbols:  $ABCA \rightarrow 010110$
- required bits:

$$\frac{\text{transmitted bits}}{\text{transmitted symbols}} = \frac{6}{4} = 1.5$$

$\Rightarrow$  optimal transmission

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## source coding

## huffman coding: tree construction 1/2

	frequency	symbol
	5	1
	7	2
sort symbols acc. to frequency	10	3
	15	4
	20	5
	45	6

	12:*	
	5:1	7:2
combine two lowest symbols into new entry (sum)	frequency	symbol
	10	3
	12	*
add new entry to list	15	4
	20	5
	45	6

repeat until only one element left in

# source coding

## huffman coding: tree construction 2/2

5:1

7:2

10:3

15:4

20:5

45:6

# source coding

## huffman coding: tree construction 2/2

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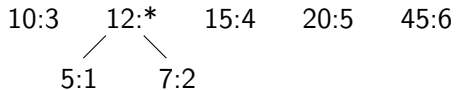
15:4

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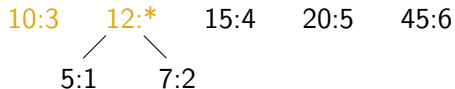
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## huffman coding: tree construction 2/2



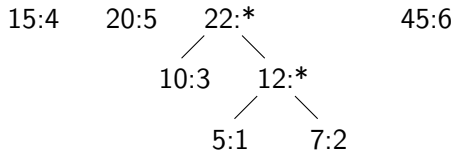
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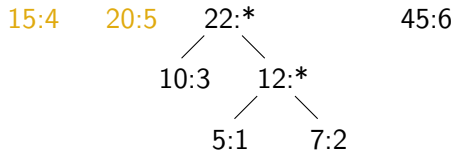
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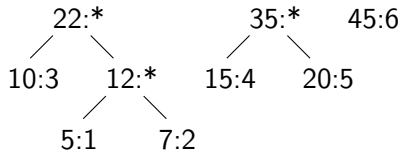
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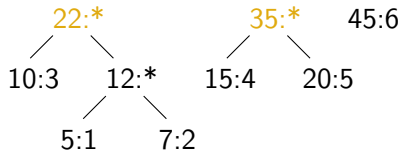
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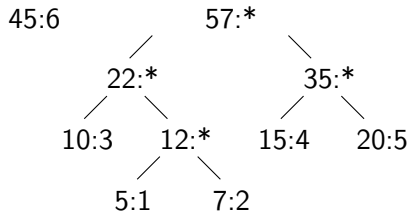
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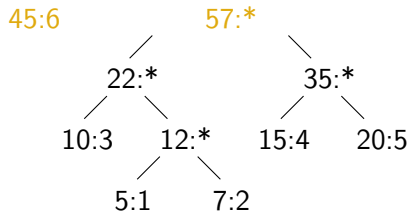
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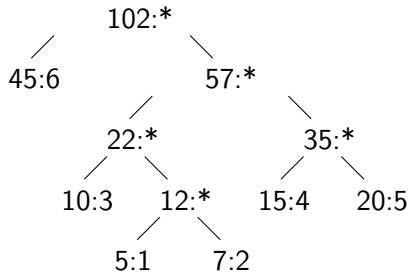
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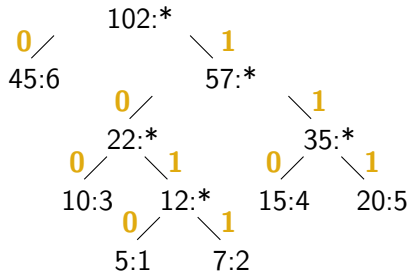
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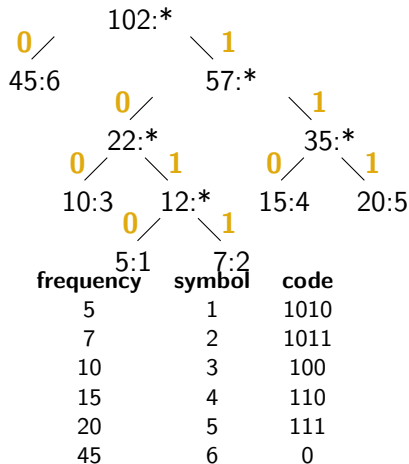
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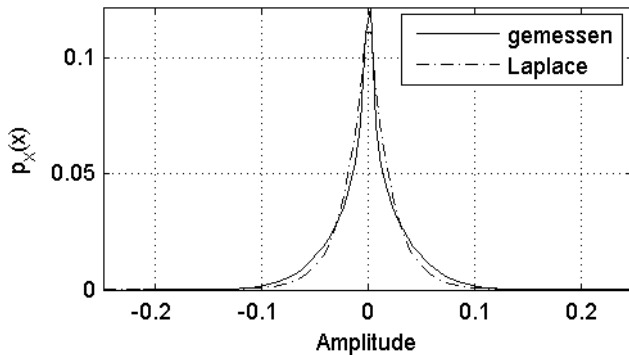


**note** no code is prefix of another code!

# source coding

## huffman coding for audio signals

- Symbole:  $2^w$
- PDF indicates probability per symbol

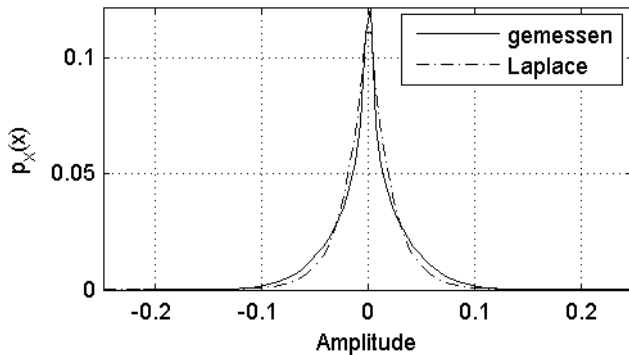




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# source coding

## arithmetic coding

- **Huffman coding is only optimal if  $p_n = \frac{1}{2^k}$**
- alternative: **arithmetic coding**
  - allows other probability distributions
  - encodes the whole sequence in one fractional number  $0.0 \leq f < 1.0$
- **principle:**
  - 1 assume initial interval of  $[0, 1[$
  - 2 assign interval segments to all symbols, e.g.  $A = [0, 0.7[, B = [0.7, 0.9[, C = [0.9, 1[$
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# source coding

## arithmetic coding: example 1/2

sequence  $ABCA$ ,  $p_A = 0.6$ ,  $p_B = 0.2$ ,  $p_C = 0.1$ ,  $p_T = 0.1$ ,  
 $A = [0, 0.6[$ ,  $B = [0.6, 0.8[$ ,  $C = [0.8, 0.9[$ ,  $T = [0.9, 1[$

### ● decoding 0.463:

- ①  $0.463 \in \text{segment 1 } (\rightarrow A)$ ,
  - set interval  $[0, 0.6[ \rightarrow \text{bounds: } 0, 0.36, 0.48, 0.54, 0.6$
- ②  $0.463 \in \text{segment 2 } (\rightarrow B)$ ,
  - set interval  $[0.36, 0.48[ \rightarrow \text{bounds: } 0.36, 0.432, 0.456, 0.468, 0.48$
- ③  $0.463 \in \text{segment 3 } (\rightarrow C)$ ,
  - set interval  $[0.456, 0.468[ \rightarrow \text{bounds: } 0.456, 0.4632, 0.4656, 0.4668, 0.468$
- ④  $0.463 \in \text{segment 1 } (\rightarrow A)$ ,
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  - set interval  $[0.456, 0.468[ \rightarrow$  bounds:  $0.456, 0.4632, 0.4656, 0.4668, 0.468$
- ④  $0.463 \in \text{segment 1 } (\rightarrow A)$ ,
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# source coding

## arithmetic coding: example 1/2

sequence  $ABCA$ ,  $p_A = 0.6$ ,  $p_B = 0.2$ ,  $p_C = 0.1$ ,  $p_T = 0.1$ ,  
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### ● decoding 0.463:

- ①  $0.463 \in \text{segment 1 } (\rightarrow A)$ ,
  - set interval  $[0, 0.6[ \rightarrow$  bounds:  $0, 0.36, 0.48, 0.54, 0.6$
- ②  $0.463 \in \text{segment 2 } (\rightarrow B)$ ,
  - set interval  $[0.36, 0.48[ \rightarrow$  bounds:  $0.36, 0.432, 0.456, 0.468, 0.48$
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# source coding

## fundamentals: linear prediction

idea: use preceding samples to estimate/predict future samples.

- **estimate the signal  $x$**

$$\hat{x}(i) = \sum_{j=1}^{\mathcal{O}} b_j \cdot x(i-j)$$

- prediction quality is measured by **power of prediction error**

$$\begin{aligned} e_P(i) &= x(i) - \hat{x}(i) \\ &= x(i) - \sum_{j=1}^{\mathcal{O}} b_j \cdot x(i-j) \end{aligned}$$

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## source coding

## fundamentals: linear prediction — first order prediction 1/2

- **prediction**  $\hat{x}(i) = b_1 \cdot x(i-1)$
- prediction error

$$\begin{aligned}\sigma_e^2 &= \mathcal{E} \left\{ (x(i) - b_1 x(i-1))^2 \right\} \\ &= \sigma_x^2 + b_1^2 \sigma_x^2 - 2b_1 r_{xx}(1) \\ &= \left( 1 + b_1^2 - 2b_1 \rho_{xx}(1) \right) \sigma_x^2\end{aligned}$$

- optimum coefficient:  $\frac{\partial \sigma_e^2}{\partial b_1} = 0$

$$\begin{aligned}2b_1 \sigma_x^2 - 2\rho_{xx}(1) \sigma_x^2 &= 0 \\ b_1 &= \rho_{xx}(1)\end{aligned}$$

- minimum prediction error power

$$\begin{aligned}\sigma_e^2 &= \left( 1 + b_1^2 - 2b_1 \rho_{xx}(1) \right) \sigma_x^2 \\ &= \left( 1 + \rho_{xx}(1)^2 - 2\rho_{xx}(1)\rho_{xx}(1) \right) \sigma_x^2 \\ &= (1 - \rho_{xx}(1)) \sigma_x^2\end{aligned}$$

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## fundamentals: linear prediction — first order prediction 2/2

$$\sigma_e^2 = (1 - \rho_{xx}(1))\sigma_x^2$$

● **observations:**

- power of prediction error always smaller or equal the power of the signal
- question: when is it equal to the signal?

● **special case:**  $b_1 = 1$ 

$$\begin{aligned}\hat{x}(i) &= x(i-1) \\ e_P &= x(i) - x(i-1) \\ \sigma_e^2 &= (1 + b_1^2 - 2b_1\rho_{xx}(1))\sigma_x^2 \\ &= 2(1 - \rho_{xx}(1))\sigma_x^2\end{aligned}$$

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# source coding

## fundamentals: linear prediction — prediction coefficients

- prediction gain depends on
  - predictor coefficients  $b_j$
  - signal
- optimal coefficients can be derived by finding minimum of prediction error

$$\frac{\partial \sigma_e^2}{\partial b_j} = 0$$

⇒ (without derivation)

$$r_{xx}(\eta) = \sum_{j=1}^{\mathcal{O}} b_{j,\text{opt}} \cdot r_{xx}(\eta - j), \quad 1 \leq \eta \leq \mathcal{O}$$

$$\begin{aligned} \mathbf{r}_{xx} &= \mathbf{R}_{xx} \cdot \mathbf{b}_{\text{opt}} \\ \mathbf{b}_{\text{opt}} &= \mathbf{R}_{xx}^{-1} \cdot \mathbf{r}_{xx} \end{aligned}$$

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## fundamentals: linear prediction — summary

- **predictor length**

- rule of thumb: the longer the predictor, the better the prediction
- can range from 10 coefficients to hundreds

- **predictor coefficient updates**

- better signal adaptation if coefficients are updated block-by-block

- **input signals**

- white noise/random processes cannot be predicted
- periodic signals may theoretically be perfectly predicted

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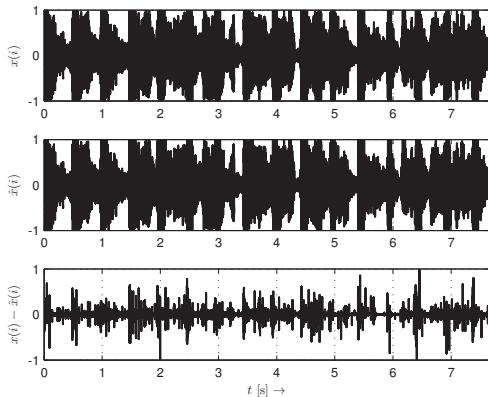
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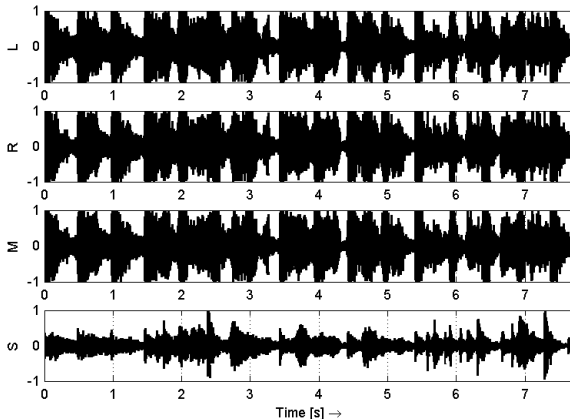
## fundamentals: linear prediction — audio example

order: 20



# source coding

## fundamentals: joint channels



$$M = \frac{L+R}{2}$$



$$S = \frac{L-R}{2}$$

$$L = M + S$$

$$R = M - S$$

# source coding

## summary

- bitrate can be reduced by removing removing redundancy and/or irrelevance
- removing redundancy:
  - entropy coding: transmit frequent symbols with shorter codes
  - linear prediction: transmit diff signal plus predictor coefficients
- removing irrelevance:
  - reduce quantization wordlength
  - see slides below