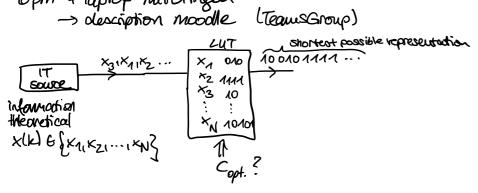
10. lecture

Freitag, 20. Oktober 2023 09:56

6pm + laptop mitteringen



Hemoryless & stacionary source model

$$\rho(\mathbf{x(k)} = x_i \mid \mathbf{x(k-1)} = x_i) = \rho(\mathbf{x(k)} =$$

Autrage codelength:
$$L := E(e(\kappa)) = \sum_{i=1}^{n} \rho(\kappa) e(\kappa)$$

$$H(x) = \begin{cases} p(x) & (d \frac{1}{p(x)}) \\ 0 \leq H(x) \leq (d N) \end{cases}$$

avoage inform. provided by source = Entropie

Shannon-Fano codes:

H(x) =
$$\underset{\times}{\text{find}} p(x) \text{ ld } p(x) = \underset{\times}{\text{find}} p(x) \text{ Tid } p(x)$$

Algo: given $p(x)$ - source distribution $\mapsto l(x) := \text{Id } p(x)$
 $\Rightarrow \text{ binary tree} \rightarrow \overline{c}(x) \rightarrow \text{ lut}$
 $\underset{\times}{\text{find}} 2^{-l(x)} \le 1 \text{ lprefix free } V$

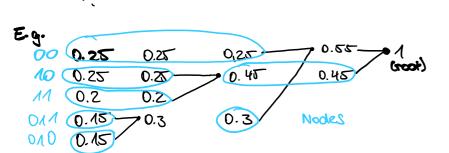
H(x) € L_{SF} € H(x) +1

Huffman Coding

$$P_1 > P_2 > ... > P_{N-1} > P_N$$

Criteria for optimality

$$e_{N-1} = e_N$$



1. two smallest 2 throwback - time window for probabilities I'm every time window calculate the code new

$$\mathbf{F}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) \qquad \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i=1} \rho(\mathbf{x}_{i}) + \frac{1}{2} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i=1} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i})$$

$$\widehat{\mathbf{F}}(\mathbf{x}_{i}) - \widehat{\mathbf{F}}(\mathbf{x}_{i}) = \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) + \sum_{j=1}^{i} \rho(\mathbf{x}_{i}) = \sum_{$$

$$2^{2-C(k)} = 2^{2} 2^{-[kd \frac{1}{p(k)}]-1} \le 2^{2} 2^{-kd \frac{1}{p(k)}-1} = 2^{2} 2^{2} 2^{-kd \frac{1}{p(k)}}$$

$$= 2^{2} 2^{2} 2^{-kd \frac{1}{p(k)}}$$

Algo: given
$$\rho(x_i)_i := 1,...,N \rightarrow \widehat{+}(x_i) = \frac{1}{2} \rho(x_i) + \frac{1}{2} \rho(x_i)$$

$$\rightarrow \rho(x_i) = \left[\log + \frac{1}{\rho(x_i)} \right] + 1 \rightarrow \left[\widehat{+}(x_i) \right]_{i=1} \rightarrow C(x_i)$$

(no binary tree, no seasoning for two minimal propability)

>> price: average L großer dafür fanter

-> small improvement in " 2" matters a lot in dataspood

Huffman 372,8 Hbps binasy tree
STE highes lasts als Huffman

=) posametric methods: know the propability distributions before