Entropy-Based Measures (II): Normalized Mutual Information (NMI)

■ Mutual information:

- Quantifies the amount of shared info between $I(C,T) = \sum_{i=1}^{r} \sum_{j=1}^{k} p_{ij} \log(\frac{p_{ij}}{p_{C_i} \cdot p_{T_j}})$ the clustering C and partitioning T
- \square Measures the dependency between the observed joint probability p_{ij} of C and T, and the expected joint probability p_{Ci} . p_{Tj} under the independence assumption
- □ When C and T are independent, $p_{ij} = p_{Ci} \cdot p_{Tj}$, I(C, T) = 0. However, there is no upper bound on the mutual information
- Normalized mutual information (NMI)

$$NMI(\mathcal{C},\mathcal{T}) = \sqrt{\frac{I(\mathcal{C},\mathcal{T})}{H(\mathcal{C})} \cdot \frac{I(\mathcal{C},\mathcal{T})}{H(\mathcal{T})}} = \frac{I(\mathcal{C},\mathcal{T})}{\sqrt{H(\mathcal{C}) \cdot H(\mathcal{T})}}$$

■ Value range of NMI: [0,1]. Value close to 1 indicates a good clustering

