

MDL

(Minimum Description Length)

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CoSCo

<http://cosco.hiit.fi/>

<http://www.mdl-research.org/>

MDL

The shortest Code is the best Model

$$P(D; M) = 2^{-L(D; M)}$$

$L(D; M)$ approximates the Kolmogorov Complexity

The Best Model

- **compresses the data**
- **describes main features in short**
- **predicts**
- **denoises**
- **...**

Classic MDL

two-part code

$$L_2(D; M) = L(M, \Theta) + L(D; M, \Theta)$$

→ redundancies

Modern MDL

NML (Normalized Maximum Likelihood)

$$P_{NML}(D; M) = \frac{P(D; M, \hat{\Theta}(D))}{\sum_{D'} P(D; M, \hat{\Theta}(D'))}$$

→ **shorter code, better model**

→ **minimax optimal**

→ **loses $\log \sum_{D'} P(D; M, \hat{\Theta}(D'))$ (“regret”) bits on any data against hind-sighted opponent**

NML: HowTo

$$P_{NML}(D; M) = \frac{P(D; M, \hat{\Theta}(D))}{\sum_{D'} P(D; M, \hat{\Theta}(D'))}$$

- no (parameter) prior needed
- use it for model class selection
- use it for clustering, prediction, denoising, ...

NML: Problems

$$P_{NML}(D; M) = \frac{P(D; M, \hat{\Theta}(D))}{\sum_{D'} P(D; M, \hat{\Theta}(D'))}$$

- regret typically incomputable
- still need to code the model class
- model class vs. parameters
- regret is a function of $|D|$

NML: Applications

$$P_{NML}(D; M) = \frac{P(D; M, \hat{\Theta}(D))}{\sum_{D'} P(D; M, \hat{\Theta}(D'))}$$

- **Image/Signal Denoising**
- **Clustering**
- **Density Estimation**
- **Bayesian Network Selection Criterion**
- **Hot Topic: Conditional/Supervised/Local NML**