

Minimax strategy for prediction with expert advice under stochastic assumptions

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Prediction with expert advice

In trials $t = 1, 2, \dots, T$:

- Algorithm predicts with $\mathbf{w}_t \in \Delta^K$.
 - Loss vector $\ell_t \in [0, 1]^K$ is revealed.
 - Algorithm incurs loss $\mathbf{w}_t \cdot \ell_t$.
-

Regret of a strategy $\omega = (\mathbf{w}_1, \dots, \mathbf{w}_T)$:

$$R = \sum_t \mathbf{w}_t \cdot \ell_t - \min_k \underbrace{\sum_{t \leq T} \ell_{t,k}}_{L_{T,k}}$$

Goal: find ω minimizing the worst-case regret over all sequences.

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Drop the minimax principle?
Drop the worst-case assumptions?

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Stochastic setting: optimal strategy for “easy” data

Assumption: Each expert $k = 1, \dots, K$ generates losses i.i.d. from a fixed distribution P_k .

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Expected regret: $R_{\text{eg}}(\omega, \mathcal{P}) = \mathbb{E}[\sum_t \mathbf{w}_t \cdot \ell_t - \min_k L_{T,k}]$

Redundancy: $R_{\text{ed}}(\omega, \mathcal{P}) = \mathbb{E}[\sum_t \mathbf{w}_t \cdot \ell_t] - \min_k \mathbb{E}[L_{T,k}]$

Excess risk: $R_{\text{isk}}(\omega, \mathcal{P}) = \mathbb{E}[\mathbf{w}_T \cdot \ell_T] - \min_k \mathbb{E}[\ell_{T,k}]$

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We give a strategy ω^* , which is minimax with respect to all three measures **simultaneously**:

$$\sup_{\mathcal{P}} R(\omega^*, \mathcal{P}) = \inf_{\omega} \sup_{\mathcal{P}} R(\omega, \mathcal{P}),$$

where R is either R_{eg} , R_{ed} , or R_{isk} .

Excess risk: $R_{\text{isk}}(\omega, \mathcal{P}) = \mathbb{E}[\mathbf{w}_T \cdot \ell_T] - \min_k \mathbb{E}[\ell_{T,k}]$