

## **EXPECTATION DEVIATION UPPER BOUND**

## Why

We bound the probability that a random variance deviates from its mean using its variance.

## Result

**Proposition 1.** Let f be a square-integrable real-valued random variable on the probability space  $(X, A, \mu)$ . Then for t > 0,

$$\mu[|f - \mathbf{E}(f)| \ge t] \le \frac{\operatorname{var} f}{t^2}.$$

*Proof.* The symbols  $|f - \mathbf{E}(f)| \ge t$  denote the set  $\{x \in X \mid |f(x) - \mathbf{E}(f)| \ge t\}$ . This set is the same as the set

$$\{x \in X \mid (f(x) - \mathbf{E}(f))^2 \ge t^2\}.$$

By using the tail measure upper bound,

$$\mu(\left\{x\in X \; \big|\; (f(x)-\mathbf{E}(f))^2 \geq t^2\right\}) \leq \frac{\mathbf{E}(f-\mathbf{E}(f))^2}{t^2}.$$

We recognize the numerator of the right hand side as the variance of f.  $\square$ 

The above is also called *Chebychev's Inequality*.

