Inequalities

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1 Inequalities

I Triangle Inequality (三角不等式)

For any normed space V with norm $\|\cdot\|$.

$$\forall a, b \in V : ||a|| + ||b|| \ge ||a + b|| \ge |||a|| - ||b|||$$

II Reverse Triangle Inequality (反三角不等式)

For any normed space V with norm $\|\cdot\|$.

$$\forall a, b \in V : ||a|| + ||b|| \ge ||a - b|| \ge |||a|| - ||b|||$$

III Jensen's Inequality (詹森不等式)

Let (Ω, μ) be a probability space, $g: \Omega \to \mathbb{R}$ be a real-valued μ -integrable function, and $\varphi: \mathbb{R} \to \mathbb{R}$ be a convex function. Then, Jensen's inequality states:

$$\int_{\Omega} \varphi \circ g(\omega) \, \mathrm{d}\mu(\omega) \ge \varphi \left(\int_{\Omega} g(\omega) \, \mathrm{d}\mu(\omega) \right)$$

Proof.

Since φ is convex, at each real number x, we have a non-empty set of subderivatives, which may be thought of as lines touching the graph of φ at x, but which are below the graph of φ at all points.

We define:

$$x_0 := \int_{\Omega} g \, \mathrm{d}\mu.$$

Because of the existence of subderivatives for convex functions, we may choose a and b such that $ax + b \le \varphi(x)$ for all real x and $ax_0 + b = \varphi(x_0)$.

But then we have that for almost all $\omega \in \Omega$:

$$\varphi(g(\omega)) \ge ag(\omega) + b$$

Since we have a probability measure, the integral is monotone with $\mu(\Omega) = 1$ so that

$$\begin{split} \int_{\Omega} \varphi \circ g(\omega) \, \mathrm{d}\mu(\omega) & \geq \int_{\Omega} a g(\omega) + b \, \mathrm{d}\mu(\omega) \\ &= a \int_{\Omega} g \, \mathrm{d}\mu + b \int_{\Omega} \, \mathrm{d}\mu \\ &= a x_0 + b = \varphi(x_0) \\ &= \varphi\left(\int_{\Omega} g \, \mathrm{d}\mu\right). \end{split}$$

IV AM-GM Inequality (算幾不等式)

$$\frac{\sum_{i=1}^{n} x_i}{n} \ge \sqrt[n]{\prod_{i=1}^{n} x_i}$$

Proof.

Lemma: Jensen's inequality.

Applying Jensen's inequality to the logarithm function, which is concave, and the arithmetic mean:

$$\log\left(\frac{\sum_{i=1}^{n} x_i}{n}\right) \ge \sum_{i=1}^{n} \frac{1}{n} \log\left(x_i\right) = \log\left(\sqrt[n]{\prod_{i=1}^{n} x_i}\right)$$

Exponentiating both sides gives the desired inequality:

$$\frac{\sum_{i=1}^{n} x_i}{n} \ge \sqrt[n]{\prod_{i=1}^{n} x_i}$$

V Cauchy-Schwarz Inequality (柯西-施瓦茨不等式)

For any $\mathbf{u}, \mathbf{v} \in \mathbb{C}^n$,

$$|\mathbf{u} \cdot \mathbf{v}| \le |\mathbf{u}| \, |\mathbf{v}|$$

and

$$|\mathbf{u} \cdot \mathbf{v}| \le |\mathbf{u}| \, |\mathbf{v}| \iff \mathbf{u}/\mathbf{v},$$

where $\mathbf{u} \cdot \mathbf{v}$ is the standard inner product defined as

$$\mathbf{u} \cdot \mathbf{v} = \sum_{i=1}^{n} u_i \overline{v_i},$$

with $\overline{v_i}$ denoting the complex conjugate of v_i .

Proof.

Consider the complex number $z = \mathbf{u} \cdot \mathbf{v}$.

Define the function

$$f(t) = |\mathbf{u} + t\mathbf{v}|^2$$

for some real number t. Then, we have

$$f(t) = (\mathbf{u} + t\mathbf{v}) \cdot \overline{(\mathbf{u} + t\mathbf{v})}$$

$$= (\mathbf{u} \cdot \overline{\mathbf{u}}) + t(\mathbf{u} \cdot \overline{\mathbf{v}}) + t(\mathbf{v} \cdot \overline{\mathbf{u}}) + t^{2}(\mathbf{v} \cdot \overline{\mathbf{v}})$$

$$= |\mathbf{u}|^{2} + 2t\Re(\mathbf{u} \cdot \overline{\mathbf{v}}) + t^{2}|\mathbf{v}|^{2}.$$

Since $f(t) \ge 0$ for all $t \in \mathbb{R}$, the quadratic equation in t must have a non-positive discriminant. The discriminant of this quadratic is:

$$\Delta = (2\Re(\mathbf{u} \cdot \bar{\mathbf{v}}))^2 - 4 \times |\mathbf{v}|^2 \times |\mathbf{u}|^2$$
$$= 4 (\Re(\mathbf{u} \cdot \bar{\mathbf{v}}))^2 - 4|\mathbf{u}|^2 |\mathbf{v}|^2.$$

We require $\Delta \leq 0$. This implies:

$$\left(\Re(\mathbf{u}\cdot\bar{\mathbf{v}})\right)^2 \leq |\mathbf{u}|^2|\mathbf{v}|^2.$$

Taking the square root of both sides and noting that $|u\cdot v| \geq \Re(u\cdot \bar{v})$:

$$|\mathbf{u} \cdot \mathbf{v}| \le |\mathbf{u}| |\mathbf{v}|$$
.

Equality $|\mathbf{u} \cdot \mathbf{v}| = |\mathbf{u}| |\mathbf{v}|$ holds if and only if the discriminant $\Delta = 0$. This happens when the quadratic equation has a double root or equivalently, when $\mathbf{u} + t\mathbf{v} = 0$ for some real t, implying \mathbf{u} is a scalar multiple of \mathbf{v} . Therefore, \mathbf{u} and \mathbf{v} are linearly dependent, meaning

u/v.