

**Theorem 0.0.1** (conditional expectation). Let  $X, Y$  be two integrable random variables on  $(\Omega, \mathcal{A}, \mathbb{P})$ , and  $\mathcal{G}, \mathcal{H}$  be two sub  $\sigma$ -algebras of  $\mathcal{A}$ . Then

1. **(taking out what is known)** If  $X$  is  $\mathcal{G}$ -measurable (or equivalently  $\sigma(X) \subseteq \mathcal{G}$ ), then  $\mathbb{E}[X|\mathcal{G}] = X$ .

2. **(linearity)** For  $a, b \in \mathbb{R}$ ,  $\mathbb{E}[aX + bY|\mathcal{G}] = a\mathbb{E}[X|\mathcal{G}] + b\mathbb{E}[Y|\mathcal{G}]$ .

3. **(tower property)** If  $\mathcal{H} \subseteq \mathcal{G}$  then

$$\mathbb{E}[\mathbb{E}[X|\mathcal{G}]|\mathcal{H}] = \mathbb{E}[X|\mathcal{H}]$$

in particular, by taking  $\mathcal{H} = \{\emptyset, \Omega\}$  to be the trivial  $\sigma$ -algebra, we have

$$\mathbb{E}[\mathbb{E}[X|\mathcal{G}]] = \mathbb{E}[X].$$

4. If  $X$  is independent of  $\mathcal{G}$  in the sense that for all  $A \in \sigma(X)$  and  $B \in \mathcal{G}$  we have  $\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$ , then

$$\mathbb{E}[X|\mathcal{G}] = \mathbb{E}[X].$$

5. If  $X$  is  $\mathcal{G}$ -measurable and  $Y$  is independent of  $\mathcal{G}$  then for any Borel function  $h : \mathbb{R}^2 \rightarrow \mathbb{R}$  we have

$$\mathbb{E}[h(X, Y)|\mathcal{G}] = H(X),$$

where  $H : \mathbb{R} \rightarrow \mathbb{R}$  is given by the formula  $H(x) = \mathbb{E}[h(x, Y)]$ . Consider, e.g.  $X$  represents the present,  $Y$  the future and  $\mathcal{G}$  the information generated by past events.

6. **(conditional Jensen's inequality)** Let  $g : \mathbb{R} \rightarrow \mathbb{R}$  be a convex function and for any  $\sigma$ -algebra  $\mathcal{G} \subseteq \mathcal{A}$ ,

$$g(\mathbb{E}[X|\mathcal{G}]) \leq \mathbb{E}[g(X)|\mathcal{G}].$$