

MARKOV KERNELS V1: PROBABILITY



1. For all $x \in X$, $A \mapsto K(x, A) \in \mathcal{P}(X)$,

► A Markov kernel $K: \mathcal{X} \times \mathcal{F} \rightarrow [0, 1]$ is a function such that

2. For all $A \in \mathcal{F}$, $x \mapsto K(x, A)$ is \mathcal{F} -measurable

► In general we will denote $K(x, dx')$ as the measure may not have a density over dx'

► We will abuse notation and use $K(x, x')$ interchangeably

- ▶ E.g. when $\mathbb{X} = \{x_1, \dots, x_n\}$ is discrete:
 - ▶ We can represent K as a $n \times n$ -dimensional square matrix with entries $K(x_i, x_j)$
 - ▶ Satisfies $K(x, x') \geq 0$ the rows sum to 1

$$\sum_{x'} K(x, x') = 1$$

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 1. For all $x \in \mathbb{X}$, $A \mapsto K(x, A) \in \mathcal{P}(\mathbb{X})$,
 2. For all $A \in \mathcal{F}$, $x \mapsto K(x, A)$ is \mathcal{F} -measurable
- ▶ In general we will denote $K(x, dx')$ as the measure may not have a density over dx'
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MARKOV KERNELS V2: ALGORITHMS