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Measure-preserving And Discrete (MAD) map

goal: approximate *univariate* discrete posterior $p(x), \ x \in \mathbb{N}$