### Quantile MDP

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### Quantile

• Quantile = Value-at-Risk (VaR)

$$Q_{\tau}(Y) = \inf \{ \lambda \in \mathbb{R} : \mathbb{P}[Y \leq \lambda] \geqslant \tau \}$$

- However, Quantile MDP: Non-econ ≠ Econ
- · Hence, they do not cite each other

### **MDP**

• State:  $x \in X$ 

• Action:  $a \in A$ 

• Reward function:  $r: X \times A \to \mathbb{R}$ 

• Transition: P(x, a, x')

Goal: Looking for the optimal decision rule



### Value of a Decision Rule

Under discounted reward criterion

$$v_{\pi}(x) = \mathbb{E}_{\pi} \left[ \sum_{t=0}^{\infty} \beta^{t} r(X_{t}, A_{t}) | X_{0} = x \right]$$

Under steady-state reward criterion

$$v_{\pi}(x) = \lim_{t \to \infty} \mathbb{E}_{\pi} \left[ r(X_t, A_t) | X_0 = x \right]$$

### Motivation from Applications

DM wants to optimize a quantile of rewards instead of their expectation

• A physician determines the optimal drug regime by maximizing the 0.10 quantile of improvement in health

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(want it to work with at least 90% probability for the patient)
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 Amazon determines the optimal cloud service by maximizing the 0.01 quantile of customers' satisfaction

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(provide service that satisfies 99% of its customers)
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 Basel Accord requires banks to hold capital reserves to cover at least their 10-day 99% VaR of their loss distribution

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(in the worst 1% of cases, losses could be worse than this number)
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# Xia and Pan (2025)

Under steady-state reward criterion and consider all RS  $\pi$ 

$$\begin{aligned} & \operatorname{VaR}^* = \sup_{\pi \in \Pi} \operatorname{VaR}^{\pi}_{\tau} \\ & \operatorname{VaR}^{\pi}_{\tau} = \inf \left\{ \lambda \in \mathbb{R} : \mathbb{P}_{\pi}[R^{\pi}_{\infty} \leq \lambda] \geq \tau \right\} \end{aligned}$$

where  $\mathbb{P}_{\pi}$  is the limiting distribution of the Markov chain under  $\pi$ 

Assume each  $\pi$  exactly has a  $\mathbb{P}_{\pi}$ , which is independent of initial state

#### **Transform**

They transforms

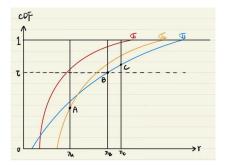
$$VaR^* = \sup_{\pi \in \Pi} VaR_{\tau}^{\pi}$$

to

$$\mathrm{VaR}^* = \min \left\{ \lambda : \min_{\sigma \in \Sigma} \left[ \mathbb{P}_{\sigma} \{ R_{\infty}^{\sigma} \leqslant \lambda \} \right] \geqslant \tau \right\}$$

where  $\Sigma$  is all DS  $\sigma$ 

$$\mathrm{VaR}^* = \min \left\{ \lambda : \min_{\sigma} \left[ \mathbb{P}_{\sigma} \{ R_{\infty}^{\sigma} \leq \lambda \} \right] \geqslant \tau \right\}$$



- Consider  $\Sigma = {\sigma_1, \sigma_2, \sigma_3}$ , the figure shows their limiting distribution of reward
- $\lambda_B$  and  $\lambda_C$  satisfy the condition since the minimum points B and C are not less than au
- $\lambda_B \leq \lambda_C$ , so it is  $VaR^*$

# Li et al. (2022)

Under discounted reward criterion

$$v(x) = \max_{\pi \in \Pi} Q_{\tau}^{\pi} \left[ \sum_{t=0}^{\infty} \beta^{t} r(X_{t}, A_{t}) | X_{0} = x \right]$$

where

- $\pi = \{\mu_0, \mu_1, \cdots \}$
- $\mu_t$  maps historical information  $h_t = (x_0, a_0, \cdots, x_{t-1}, a_{t-1}, x_t)$  to a feasible action  $a_t \in A$

Challenge: it is non-additive and non-Markovian (or, in Econ literature, not dynamically consistent)

## de Castro and Galvao (2019)

$$\begin{split} &V_{1}^{Q_{\tau}}(h,x,z^{t})\\ &=u(x_{1}^{h},x_{2}^{h},z_{1})+\beta Q_{\tau}\big[u(x_{2}^{h},x_{3}^{h},z_{2})+\beta Q_{\tau}\big[V_{2}^{Q_{\tau}}(h,x,z^{t})|Z_{2}=z_{2}\big]|Z_{1}=z\big]\\ &=Q_{\tau}\big[Q_{\tau}\big[u(x_{1}^{h},x_{2}^{h},z_{1})+\beta u(x_{2}^{h},x_{3}^{h},z_{2})+\beta^{2}V_{2}^{Q_{\tau}}(h,x,z^{t})|Z_{2}=z_{2}\big]|Z_{1}=z\big]\\ &=Q_{\tau}\bigg[Q_{\tau}\bigg[Q_{\tau}\bigg[\sum_{t=1}^{3}\beta^{t-1}u(x_{t}^{h},x_{t+1}^{h},z_{t})+\beta^{3}V_{3}^{Q_{\tau}}(h,x,z^{t})\Big|Z_{3}=z_{3}\bigg]\Big|Z_{2}=z_{2}\bigg]\Big|Z_{1}=z\bigg]\\ &=Q_{\tau}\bigg[\cdots Q_{\tau}\bigg[\sum_{t=1}^{n}\beta^{t-1}u(x_{t}^{h},x_{t+1}^{h},z_{t})+\beta^{n}V_{n}^{Q_{\tau}}(h,x,z^{t})\Big|Z_{n}=z_{n}\bigg]\Big|\cdots\Big|Z_{1}=z\bigg], \end{split}$$

Take quantile each time!

### Their Contribution

Dynamic Consistency

$$v(x) = \sup_{a \in \Gamma(x)} \left\{ r(x, a) + \beta Q_{\tau} \left[ v(x') | (x, a) \right] \right\}$$

- Advantages of quantile preferences
  - Capture heterogeneity ( $\tau$  as a parameter)
  - Separate risk aversion and elasticity of intertemporal substitution

# de Castro et al. (2025)

$$(Tv)(x, z) = \sup_{a \in \Gamma(x, z)} \{ r(x, z, a) + \beta Q_{\tau} [v(f(x, a, z'), z')|z] \}$$

where  $z' \sim P(z, \cdot)$ 

If the following conditions hold

- 1. Z is either connected or finite
- 2.  $\forall z \in \mathsf{Z} \text{ and } \varepsilon \in (0,1), \exists \mathsf{compact} \ B \subset \mathsf{Z} \text{ such that } P(z,B) > 1 \varepsilon$
- 3.  $\forall$  compact  $B \subset \mathsf{Z}$ , the map  $z \mapsto P(z, B)$  is continuous
- 4.  $\forall$  nonempty and open  $A \subset Z$  and  $z \in Z$ , P(z, A) > 0

then  $Q_{\tau}$  is a self-map on  $bc(X \times Z)$ 

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