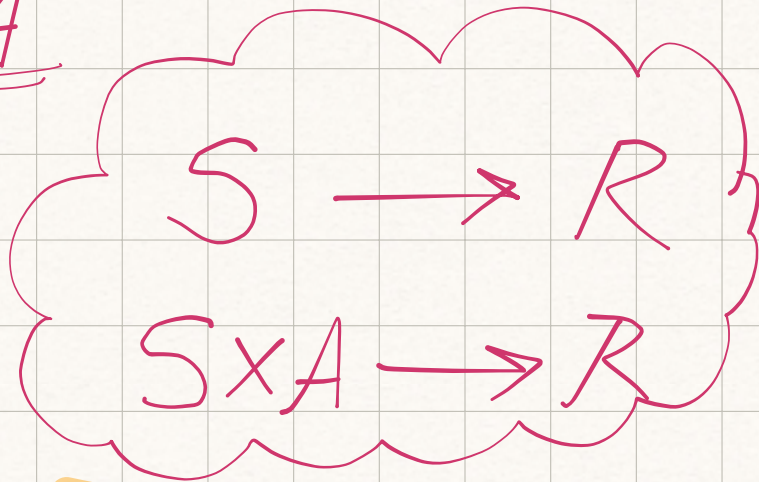


Generalizaciones del MDP

- 1) Recompensas Estado-Acción $r(s, a)$
- 2) Horizonte Finito (T) T
- 3) Sistemas Dinámicos Lineales
 - LQR
 - Basador Modelo

i) Recompensas Estado-Acción

IVA



$$S_0 \xrightarrow{a_0} S_1 \xrightarrow{a_1} S_2 \dots$$

$$R(S_0, a_0) + \gamma R(S_1, a_1) + \gamma^2 R(S_2, a_2)$$

Ec. Bellman

$$V^*(s) = \max_a [R(s, a) + \gamma \sum_s P_{sa}(s') V^*(s')]$$

les récompenses dépendent de l'action

$$W^*(s) = \arg \max_a R(s, \underline{a}) + \gamma \sum P_{sa}(s') V^*(s')$$

2^{de} Généralisation:

Horizon Finito

$$\text{MDP} \{S, A, \{P_{sa}\}, \underline{I}, R\}$$

(sin γ)

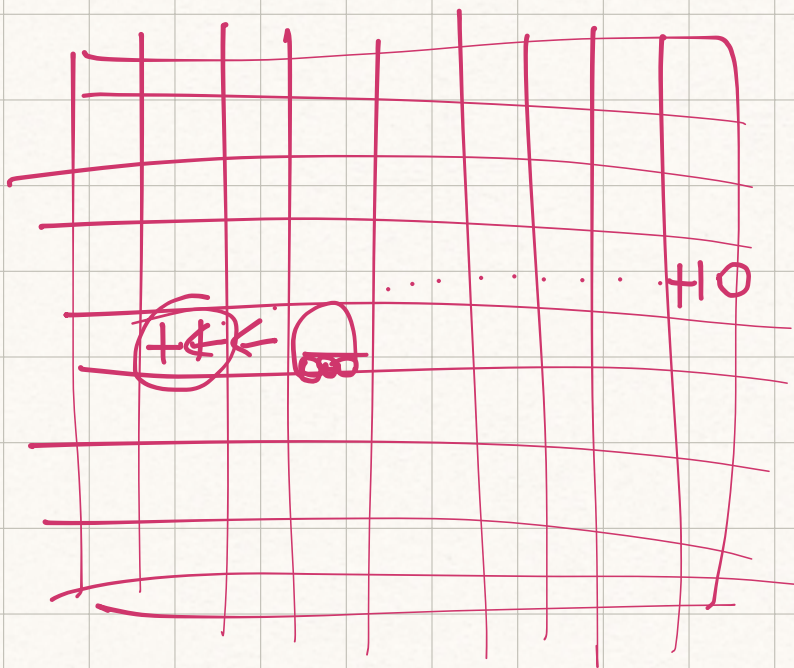
Depends la Gen 1^{re}

$$\underbrace{R(s_0, a_0) + R(s_1, a_1) + \dots + R(s_T, a_T)}_{\text{ganancia Total}} \rightarrow$$

$$\uparrow E[\text{ }]$$

maximiza

Ev vs E

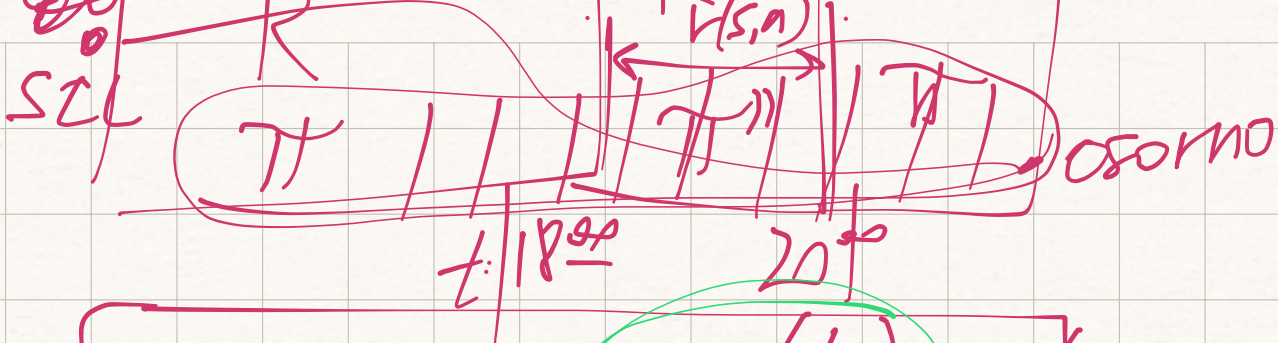


La acción óptima depende del reloj del robot.

$\pi_T^*(s) \rightarrow$ No Estacionaria
 (cambia con el tiempo)

$\pi^N(s) \rightarrow$ Estacionaria

$\pi^N(s) \rightarrow$ Estacionaria



$$S_{t+1} \sim P_{S_t a_t}^{(\underline{t})}$$

$$R_{(s, a)}^{(\underline{t})}$$

Example: Avión con combustible

$$V_t^*(s) = E \left[R(s_t, a_t) + R(s_{t+1}, a_{t+1}) + \dots + R(s_T, a_T) / \pi, s_0 = s \right]$$

$$V_t^*(s) = \max_a R(s, a) + \sum_{s'} P_{sa}^{(\underline{t})}(s') V_{t+1}^*(s')$$

$$\pi^*(s) = \arg \max_a [\dots]$$

$$V_0 \rightarrow 0 \text{ for } V_0$$

$$V_t^*(s) = \max_a R(s, a)$$

Programación Dinámica

3) Sist. Dinámicos Lineales

$$\Rightarrow MDP \{S, A, P_{sa}, T, R\}$$

$$\Rightarrow S = \mathbb{R}^n$$

$$\Rightarrow A = \mathbb{R}^d$$

$$\Rightarrow P_{sa} \rightarrow \boxed{S_{t+1} = A^{(t)} S_t + B^{(t)} a_t + \underbrace{W_t}_{\text{noise}}}$$

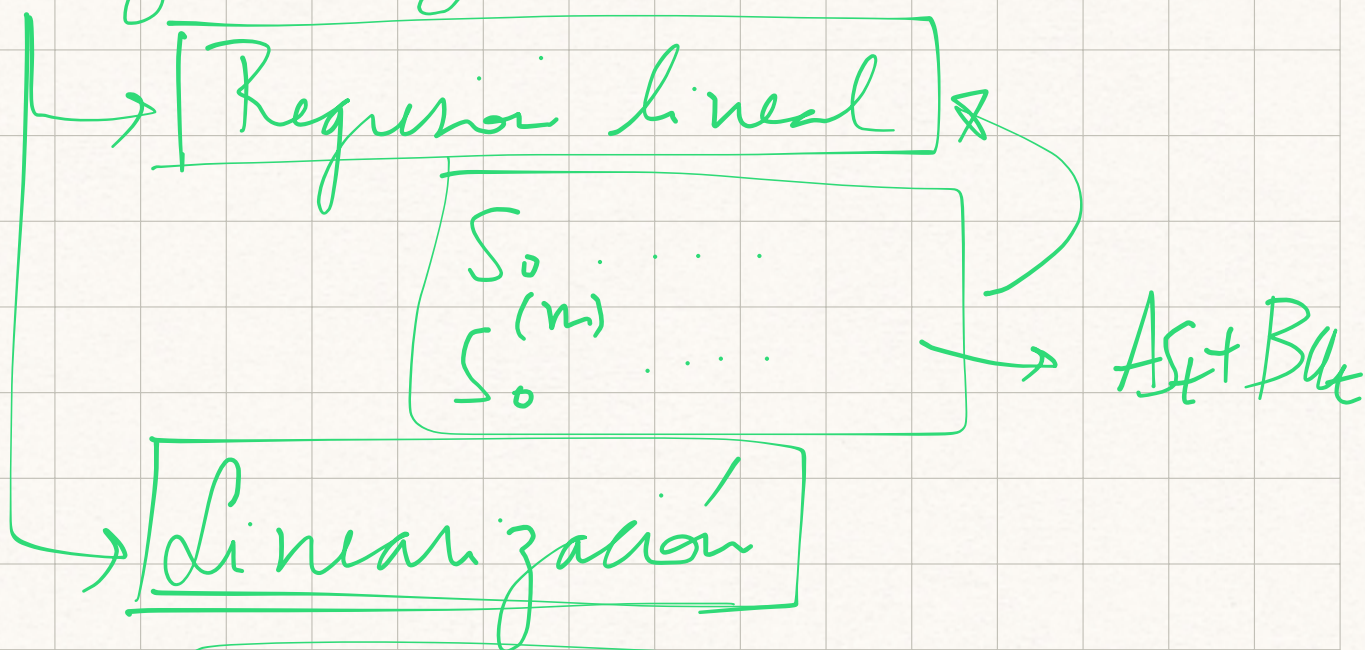
$$\Rightarrow W_t \sim \mathcal{N}(\underline{0}, \underline{\Sigma_w}) \quad \dots \underline{\text{I.i.d.}}$$

\Rightarrow Recompensas Cuadráticas

$$R(s, a) = - (S^T U_s S + a^T W_a a)$$

$$R(s, a) = -(\|s\|^2 + \|a\|^2)$$

Cómo se encuentran estas matrices A y B?



$$\underline{s}_{t+1} = f(s_t, a_t)$$

Cart Pole

$$\begin{pmatrix} x_{t+1} \\ \dot{x}_{t+1} \\ \theta_{t+1} \\ \dot{\theta}_{t+1} \end{pmatrix} = f \left(\begin{pmatrix} x_t \\ \dot{x}_t \\ \theta_t \\ \dot{\theta}_t \end{pmatrix}, a_t \right)$$

Series de Taylor

$$s_{t+1} \uparrow, f$$



$$s_{t+1} = f(s_t)$$

$$f(s_t) \approx f(\bar{s}_t) + f'(\bar{s}_t)(s_t - \bar{s}_t)$$

also
good

$$s_{t+1} = f(s_t, a_t)$$

$$s_{t+1} \approx \left[f(\bar{s}_t, \bar{a}_t) \right.$$

$$+ (\nabla_s f(\bar{s}_t, \bar{a}_t)(s_t - \bar{s}_t)$$

$$\left. + (\nabla_a f(\bar{s}_t, \bar{a}_t)(a_t - \bar{a}_t) \right]$$

Subir Webinars, etc material
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