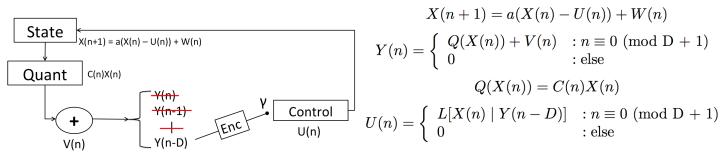
New Cost Function: Penalizing Large Control Power

Leah Dickstein

December 6, 2015

System:



New Cost Function - No V(n)

$$\min \mathbb{E}[x^2(n+1)] + \sum_{k=1}^n u^2(k)$$

$$\min \mathbb{E}[a(x(n) - u(n)) + w(n))^2] + \sum_{k=1}^n u^2(k)$$

$$= \min \mathbb{E}[(a(x(n) - \alpha y(n)) + w(n))^2] + \sum_{k=1}^n \alpha^2 y^2(k)$$

$$= \min \mathbb{E}[(a(x(n) - \alpha c(n)x(n)) + w(n))^2] + \sum_{k=1}^n \alpha^2 c^2(k)x^2(k)$$

$$= \min \mathbb{E}[a^2(1 - \alpha c(n))^2 x^2(n)] + \sigma_w^2 + \sum_{k=1}^n \alpha^2 u^2(k)$$

$$= \min \mathbb{E}[a^2(1 - \alpha c(n))^2 x^2(n)] + \sigma_w^2 + \sum_{k=1}^n \alpha^2 u^2(k) = \mathbb{E}[2a^2(1 - \alpha c(n))(-c(n))x^2(n)] + \sum_{k=1}^n 2\alpha u^2(k) = 0$$

$$- a^2(\mu_c \sigma_{x(n)}^2 - \alpha(\mu_c^2 + \sigma_c^2)\sigma_{x(n)}^2) + \sum_{k=1}^n \alpha u^2(k) = 0$$

$$\alpha(a^2(\mu_c^2 + \sigma_c^2)\sigma_{x(n)}^2 + \sum_{k=1}^n u^2(k)) = a^2\mu_c \sigma_{x(n)}^2$$

$$\alpha(n) = \frac{a^2\mu_c \sigma_{x(n)}^2}{a^2(\mu_c^2 + \sigma_c^2)\sigma_{x(n)}^2 + \sum_{k=1}^n u^2(k)}$$

This is easy to implement, and my next step is to implement this. My worry is that this isn't a closed solution like the previous α calculations. As time passes, α will change every time a new control is implemented—which is fine, but we should note the new complication. Does this answer make sense? Should I go ahead and implement this?

Original Cost Function – No V(n)

$$\min \mathbb{E}[x^{2}(n+1)]$$

$$u(n) = \alpha y(n) = \alpha c(n)x(n)$$

$$\min \mathbb{E}[(a(x(n) - u(n)) + w(n))^{2}]$$

$$= \min \mathbb{E}[(a(x(n) - \alpha y(n)) + w(n))^{2}]$$

$$= \min \mathbb{E}[(a(x(n) - \alpha c(n)x(n)) + w(n))^{2}]$$

$$= \min \mathbb{E}[a^{2}(1 - \alpha c(n))^{2}x^{2}(n)] + \sigma_{w}^{2}$$

$$\frac{d}{d\alpha}\mathbb{E}[a^{2}(1 - \alpha c(n))^{2}x^{2}(n)] + \sigma_{w}^{2} = \mathbb{E}[2a^{2}(1 - \alpha c(n))(-c(n))x^{2}(n)] = 0$$

$$\mathbb{E}[c(n)x^{2}(n) - \alpha c^{2}(n)x^{2}(n)] = 0$$

$$\mu_{c}\sigma_{x(n)}^{2} = \alpha(\mu_{c}^{2} + \sigma_{c}^{2})\sigma_{x(n)}^{2}$$

$$\alpha = \frac{\mu_{c}}{\mu_{c}^{2} + \sigma_{c}^{2}}$$

 $\mu_c^2 + \sigma_c^2$ These are the results we expect, from Gireeja's Noncoherence Paper.

Original Cost Function

$$\min \mathbb{E}[x^2(n+1)]$$

$$\begin{aligned} & \min \ \mathbb{E}[(a(1-\alpha c(n))x(n)-a\alpha v(n)+w(n))^2] \\ & = \min \ \mathbb{E}[a^2(1-\alpha c(n))^2x^2(n)+\mathbb{E}[a^2\alpha^2v^2(n)]+\sigma_w^2 \\ & \frac{d}{d\alpha}\mathbb{E}[a^2(1-\alpha c(n))^2x^2(n)+\mathbb{E}[a^2\alpha^2v^2(n)]+\sigma_w^2 = \mathbb{E}[-2a^2(1-\alpha c(n))c(n)x^2(n)]+2a^2\alpha\sigma_v^2 = 0 \\ & = -\mu_c\sigma_{x(n)}^2 + \alpha(\mu_c^2+\sigma_c^2)\sigma_{x(n)}^2 + \alpha\sigma_v^2 = 0 \end{aligned}$$

$$\alpha(n) = \frac{\mu_c \sigma_{x(n)}^2}{(\mu_c^2 + \sigma_c^2)\sigma_{x(n)}^2 + \sigma_v^2}$$

These are the results we expect from the LLSE theorem: $\alpha = \frac{cov(X,Y)}{var(Y)} = \frac{\mathbb{E}[XY]}{\mathbb{E}[Y^2]} \leftarrow \text{assuming } \mathbb{E}[X] = \mathbb{E}[Y] = 0$