

# Expected Improvement Exploration Heuristic for PILCO

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Let the parameterisation of the previous rollout be  $\mathbf{r}$ , and the cumulative-cost distribution given the previous rollout be:

$$\mathcal{C}^{\mathbf{r}} \sim \mathcal{N}(\mu_{\mathbf{r}}, \sigma_{\mathbf{r}}^2) \quad (1)$$

We would like to choose a new parameterisation,  $\theta$ , in such a way that it optimises the *expected improvement* (E.I.) of the cumulative-cost. Since we care about low costs, ‘improvement’ means a decrease in cost. For arbitrary  $\theta$  we have cumulative-cost distribution  $\mathcal{C}^{\theta} \sim \mathcal{N}(\mu_{\theta}, \sigma_{\theta}^2)$ . What is the probability  $P(\mathcal{C}^{\theta} < \mathcal{C}^{\mathbf{r}})$ ? Let  $\Delta\mathcal{C} \doteq \mathcal{C}^{\theta} - \mathcal{C}^{\mathbf{r}}$ . Note:

$$\Delta\mathcal{C} \sim \mathcal{N}(\mu_{\theta} - \mu_{\mathbf{r}}, \sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2 - 2\mathbf{c}) \quad (2)$$

where  $\mathbf{c}$  is the covariance between  $\mathcal{C}^{\theta}$  and  $\mathcal{C}^{\mathbf{r}}$ . Let’s assume (approximate) that  $\mathbf{c} = 0$ , to make life simpler. So now the expected improvement, by changing parameterisation from  $\mathbf{r}$  to  $\theta$  is:

$$\text{E.I.} = \int_{-\infty}^0 x \mathcal{N}(x; \mu_{\theta} - \mu_{\mathbf{r}}, \sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2) dx \quad (3)$$

$$= \Phi(-z)(\mu_{\theta} - \mu_{\mathbf{r}}) - \phi(z)\sqrt{\sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2} \quad (4)$$

where  $\phi(\cdot)$  is the standard normal distribution,  $\Phi(\cdot)$  its cumulative standard normal function, and  $z = \frac{\mu_{\theta} - \mu_{\mathbf{r}}}{\sqrt{\sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2}}$ . In this case the E.I. is our loss function.

$$L = \Phi(-z)(\mu_{\theta} - \mu_{\mathbf{r}}) - \phi(z)\sqrt{\sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2} \quad (5)$$

With gradients:

$$\frac{dL}{d\mu_{\theta}} = -\frac{\partial z}{\partial \mu_{\theta}} \phi(z)(\mu_{\theta} - \mu_{\mathbf{r}}) + \Phi(-z) + \frac{\partial z}{\partial \mu_{\theta}} z \phi(z) \sqrt{\sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2} \quad (6)$$

$$= \Phi(-z) \quad (7)$$

$$\begin{aligned} \frac{dL}{d\sigma_{\theta}^2} &= -\frac{\partial z}{\partial \sigma_{\theta}^2} \phi(z)(\mu_{\theta} - \mu_{\mathbf{r}}) + \frac{\partial z}{\partial \sigma_{\theta}^2} z \phi(z) \sqrt{\sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2} - \frac{\phi(z)}{2\sqrt{\sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2}} \\ &= -\frac{\phi(z)}{2\sqrt{\sigma_{\theta}^2 + \sigma_{\mathbf{r}}^2}} \end{aligned} \quad (8)$$