

# ME8135 — Assignment 1.1 Solution

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**1. Part (f):** Please visit [this link](#) for the script that produces the results documented in this report.

In part (d) we analytically computed the covariance of  $\mathbf{y}$ ,  $\Sigma_y$ , to be:

$$\Sigma_y = \mathbf{J}\Sigma\mathbf{J}^T = \begin{bmatrix} \cos\theta & -\rho\sin\theta \\ \sin\theta & \rho\cos\theta \end{bmatrix} \begin{bmatrix} \sigma_{\rho\rho}^2 & \sigma_{\rho\theta}^2 \\ \sigma_{\rho\theta}^2 & \sigma_{\theta\theta}^2 \end{bmatrix} \begin{bmatrix} \cos\theta & \sin\theta \\ -\rho\sin\theta & \rho\cos\theta \end{bmatrix} \quad (1)$$

Now given 4 different scenarios, with varying  $\Sigma$ , we wish to simulate our model using the Monte Carlo simulation and describe our observations.

Our scenarios are defined by the following equations:

$$\mathbf{x} = \begin{bmatrix} 1\text{m} \\ 0.5^\circ \end{bmatrix}, \Sigma = \begin{bmatrix} 0.01 & 0 \\ 0 & 0.005 \end{bmatrix} = \begin{bmatrix} \sigma_{\rho\rho}^2 & \sigma_{\rho\theta}^2 \\ \sigma_{\rho\theta}^2 & \sigma_{\theta\theta}^2 \end{bmatrix} \quad (2)$$

$$\mathbf{x} = \begin{bmatrix} 1\text{m} \\ 0.5^\circ \end{bmatrix}, \Sigma = \begin{bmatrix} 0.01 & 0 \\ 0 & 0.1 \end{bmatrix} \quad (3)$$

$$\mathbf{x} = \begin{bmatrix} 1\text{m} \\ 0.5^\circ \end{bmatrix}, \Sigma = \begin{bmatrix} 0.01 & 0 \\ 0 & 0.5 \end{bmatrix} \quad (4)$$

$$\mathbf{x} = \begin{bmatrix} 1\text{m} \\ 0.5^\circ \end{bmatrix}, \Sigma = \begin{bmatrix} 0.01 & 0 \\ 0 & 1 \end{bmatrix} \quad (5)$$

In the above scenarios, we are only varying  $\sigma_{\theta\theta}^2$  element of  $\Sigma$ . Initially, we use  $\sigma_{\theta\theta}^2 = 0.005$  to produce the results shown below:

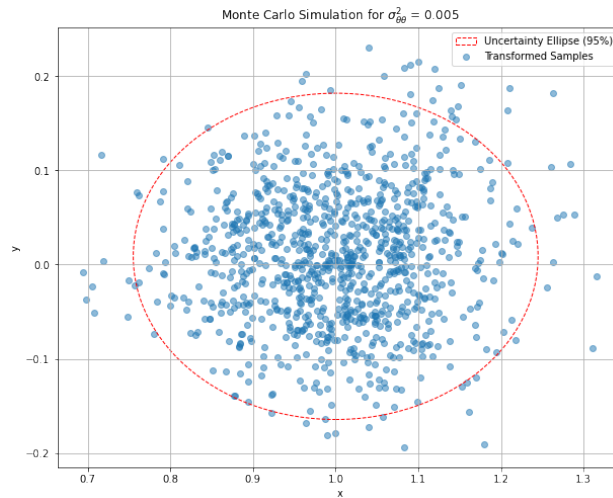


Figure 1: Monte Carlo simulation results for 2.

The simulation results in Figure 1 can be interpreted as follows:

- **Spread of samples in the  $y$  direction:** The transformed samples in the  $y$  direction,  $(\rho \sin \theta)$ , have a relatively small spread. Since  $\sigma_{\theta\theta}^2$  represents the variance of  $\theta$ , a small value indicates less variability in the angular direction. Consequently, the samples in the  $y$  direction cover a narrow range ( $\approx$  between -0.2 and 0.2).
- **Shape of the uncertainty ellipse:** The uncertainty ellipse, representing the 95% confidence region, has a smaller vertical dimension compared to the horizontal dimension. This indicates a lower uncertainty in the  $y$  direction.
- **Spread of samples in the  $x$  direction:** The transformed samples in the  $x$  direction,  $(\rho \cos \theta)$ , are influenced by  $\sigma_{\rho\rho}^2$  and independent of the value of  $\sigma_{\theta\theta}^2$ . The samples in the  $x$  direction cover a wider range compared to the ones in the  $y$  direction ( $\approx$  between 0.7 and 1.3).

In the second scenario we set  $\sigma_{\theta\theta}^2 = 0.1$  to produce the results shown below:

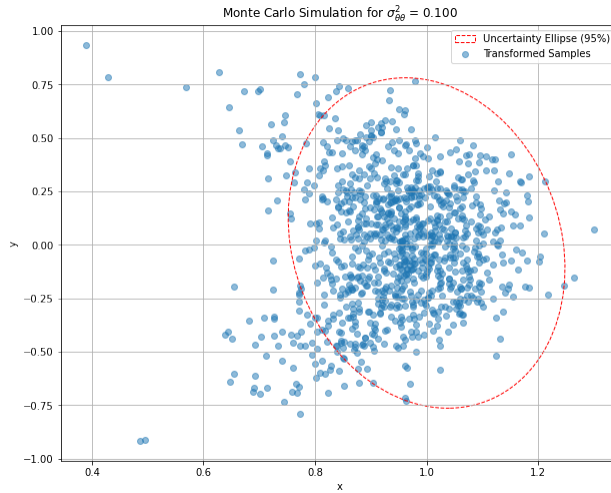


Figure 2: Monte Carlo simulation results for 3.

The simulation results in Figure 2 can be interpreted as follows:

- **Spread of samples in the  $y$  direction:** Increasing  $\sigma_{\theta\theta}^2$  to 0.1 introduces more variance in the  $\theta$  direction. As a result, the transformed samples in the  $y$  direction, which depends on  $\rho \sin \theta$ , appear to have a larger spread compared to before. Consequently, the samples in the  $y$  direction cover a wider range than in the previous scenario ( $\approx$  between -1.0 and 1.0).
- **Shape of the uncertainty ellipse:** The uncertainty ellipse is observed to be rotated counter-clockwise. The general shape of the ellipse is also growing in the vertical direction to account for the spread of samples in the  $y$  direction.
- **Spread of samples in the  $x$  direction:** The spread of the samples in the  $x$  direction, which depends on  $\rho \cos \theta$ , has reduced. The spread in the  $x$  direction covers a narrower range compared to previous scenario ( $\approx$  between -0.4 and 1.2).

In the third scenario we set  $\sigma_{\theta\theta}^2 = 0.5$  to produce the results shown in the next page:

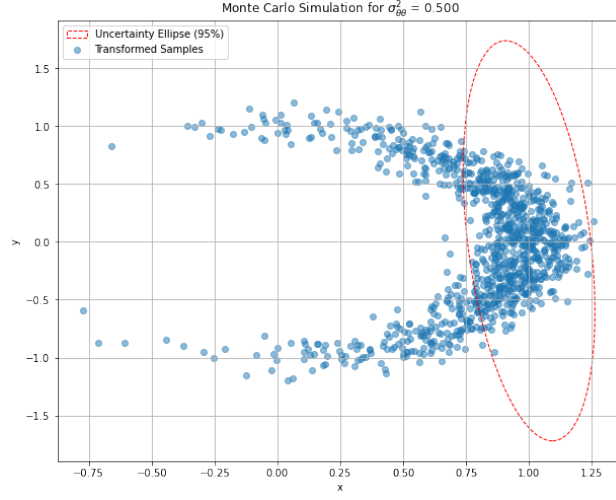


Figure 3: Monte Carlo simulation results for 4.

The simulation results in Figure 3 can be interpreted as follows:

- **Spread of samples in the  $y$  direction:** Increasing  $\sigma_{\theta\theta}^2$  to 0.5 introduces even more variance in the  $\theta$  direction. As a result, the transformed samples in the  $y$  direction demonstrate another spread. Consequently, the samples in the  $y$  direction now cover a wider range than in the previous scenario ( $\approx$  between -1.1 and 1.1).
- **Shape of the uncertainty ellipse:** The uncertainty ellipse appears to be rotated and narrowed. The general shape of the ellipse is also changing to account for the spread of samples in the  $y$  direction which makes it to be appeared more stretched.
- **Spread of samples in the  $x$  direction:** The spread of the samples in the  $x$  direction appear to have grown to mimic the spread of the first scenario ( $\approx$  between -0.75 and 1.25).

In the last scenario we set  $\sigma_{\theta\theta}^2 = 1$  to produce the results shown below:

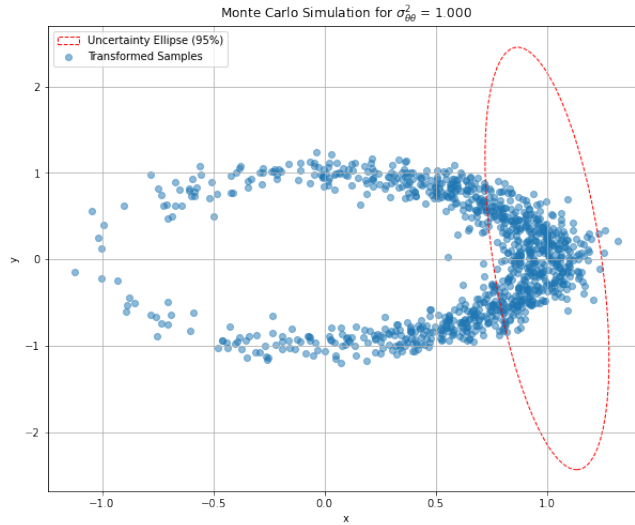


Figure 4: Monte Carlo simulation results for 5.

The simulation results in [Figure 4](#) can be interpreted as follows:

- **Spread of samples in the  $y$  direction:** Increasing  $\sigma_{\theta\theta}^2$  to 1.0 shows that the transformed samples in the  $y$  direction have more or less stopped spreading. Consequently, the samples in the  $y$  direction still cover the same range than in the previous scenario ( $\approx$  between -1.2 and 1.2).
- **Shape of the uncertainty ellipse:** The uncertainty ellipse appears to be rotated counter-clockwise and stretched when compared to the previous scenario.
- **Spread of samples in the  $x$  direction:** The spread of the samples in the  $x$  direction appear to have grown, now covering an even larger range than the previous scenario ( $\approx$  between -1.2 and 1.25).

In conclusion, smaller values of  $\sigma_{\theta\theta}^2$  correspond to smaller angular uncertainties. Increasing  $\sigma_{\theta\theta}^2$  has the reverse effect. The orientation of the uncertainty ellipse is dictated by the `angle` parameter in the script, and this angle was determined from the eigenvectors of the covariance matrix of  $\mathbf{y}$ , and the covariance matrix of  $\mathbf{y}$  depends on the covariance matrix of  $\mathbf{x}$  and the Jacobian matrix  $\mathbf{J}$ . Since  $\sigma_{\theta\theta}^2$  is stored in the covariance matrix of  $\mathbf{x}$ , changing this value affects the orientation of the uncertainty ellipse. This concept is observable from the generated simulation plots.