## 1 Online Updating of Gauss Distribution

This is implemented in goMultiGaussPDF.

$$p(x) = \frac{1}{\sqrt{\det \Sigma} (2\pi)^{\frac{n}{2}}} \exp\left(-\frac{1}{2}(x-\mu)^{\top} \Sigma^{-1}(x-\mu)\right)$$

We want to learn the covariance matrix  $\Sigma$  and mean  $\mu$ .

$$\Sigma = \frac{1}{N} \sum_{i=1}^{N} (v_i - \mu) \cdot (v_i - \mu)^{\top}$$

$$= \frac{1}{N} \sum_{i=1}^{N} (v_i - \mu) \cdot (v_i^{\top} - \mu^{\top})$$

$$= \frac{1}{N} \left[ \sum_{i=1}^{N} v_i v_i^{\top} - 2\mu \sum_{i=1}^{N} v_i^{\top} \right] + \mu \mu^{\top}$$

$$= \frac{1}{N} \left[ \sum_{i=1}^{N} v_i v_i^{\top} \right] - \mu \mu^{\top}$$

Which means we can store  $S:=\frac{1}{N}\sum_{i=1}^N v_iv_i^\top,\,\mu,$  and N and then update with a new vector v

$$\begin{array}{rcl} \mu_{k+1} & = & \frac{N_k}{N_k+1} \mu_k + \frac{v}{N_k+1} \\ S_{k+1} & = & \frac{N_k}{N_k+1} S_k + \frac{1}{N_k+1} v v^\top \\ N_{k+1} & = & N_k+1 \end{array}$$

with  $\mu_0 := 0, \mu_0 \in \mathbb{R}^n$ ,  $S_0 := 0, S_0 \in \mathbb{R}^{n \times n}$ ,  $N_0 := 0, N_0 \in \mathbb{N}^+$ .