## The Bayes Filter

System equation	$\boldsymbol{x}_k = F_{k-1} \boldsymbol{x}_{k-1} + \boldsymbol{\epsilon}_{k-1}$ $\boldsymbol{\epsilon}_k \sim N(\boldsymbol{0}, \Sigma_{\boldsymbol{\epsilon},k})$
Observation equation	$\boldsymbol{b}_{k} = A_{k}\boldsymbol{x}_{k} + \boldsymbol{e}_{k}$ $\boldsymbol{e}_{k} \sim N(\boldsymbol{0}, \boldsymbol{\Sigma}_{e,k})$
Initial conditions Other conditions	$E\{\boldsymbol{x}_0\} = \hat{\boldsymbol{x}}_0$ $E\{(\boldsymbol{x}_0 - \hat{\boldsymbol{x}}_{0 0})(\boldsymbol{x}_0 - \hat{\boldsymbol{x}}_{0 0})^T\} = P_{0 0}$ $E\{\boldsymbol{\epsilon}_k \boldsymbol{e}_j^T\} = 0,  \text{for all } k, j$
Prediction of state vector Prediction of covariance matrix Covariance matrix trix for updating	$\hat{\mathbf{x}}_{k k-1} = F_{k-1}\hat{\mathbf{x}}_{k-1 k-1}$ $P_{k k-1} = F_{k-1}P_{k-1 k-1}F_{k-1}^{T} + \Sigma_{\epsilon,k}$ $P_{k k} = (P_{k k-1}^{-1} + A_{k}^{T}\Sigma_{e,k}^{-1}A_{k})^{-1}$
Kalman gain matrix Updating of state vector	$K_k = P_{k k} A_k^{T} \Sigma_{e,k}^{-1}$ $\hat{\boldsymbol{x}}_{k k} = \hat{\boldsymbol{x}}_{k k-1} + K_k (\boldsymbol{b}_k - A_k \hat{\boldsymbol{x}}_{k k-1})$