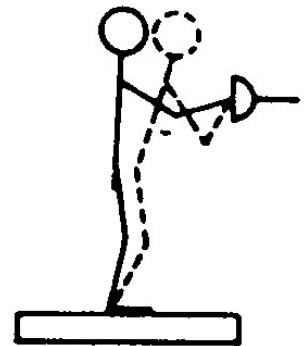
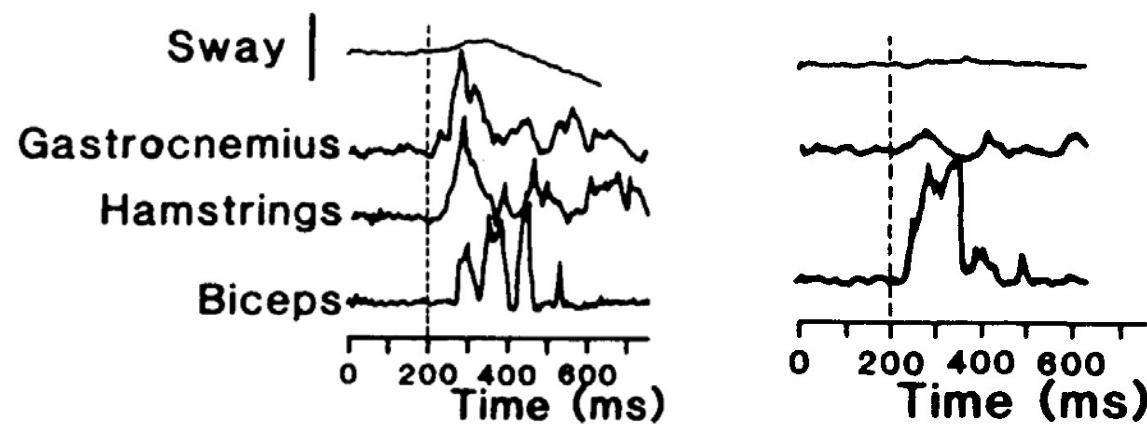
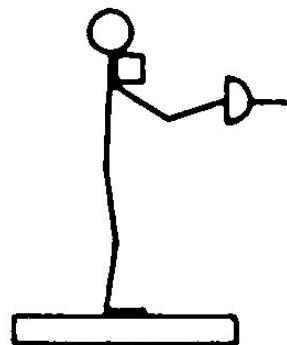
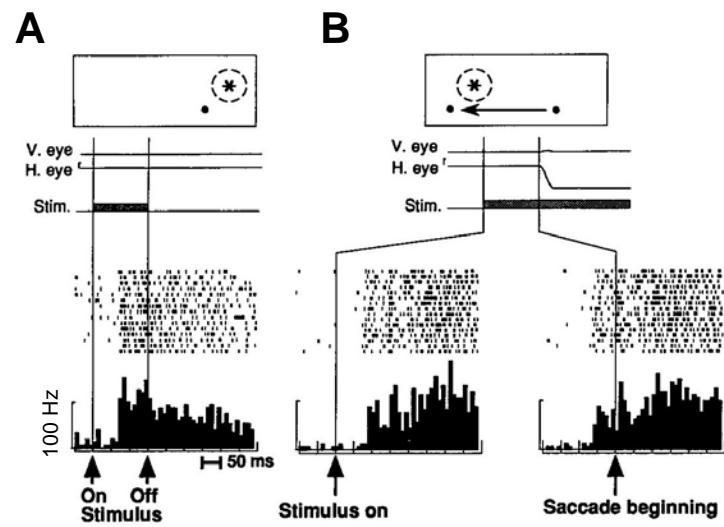


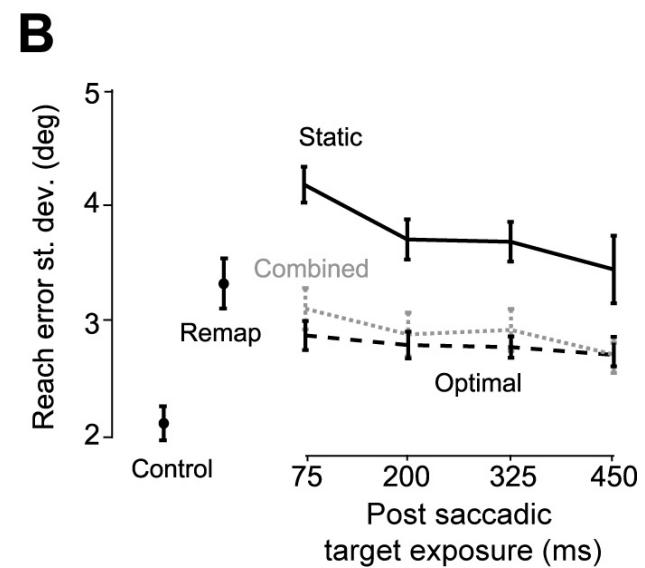
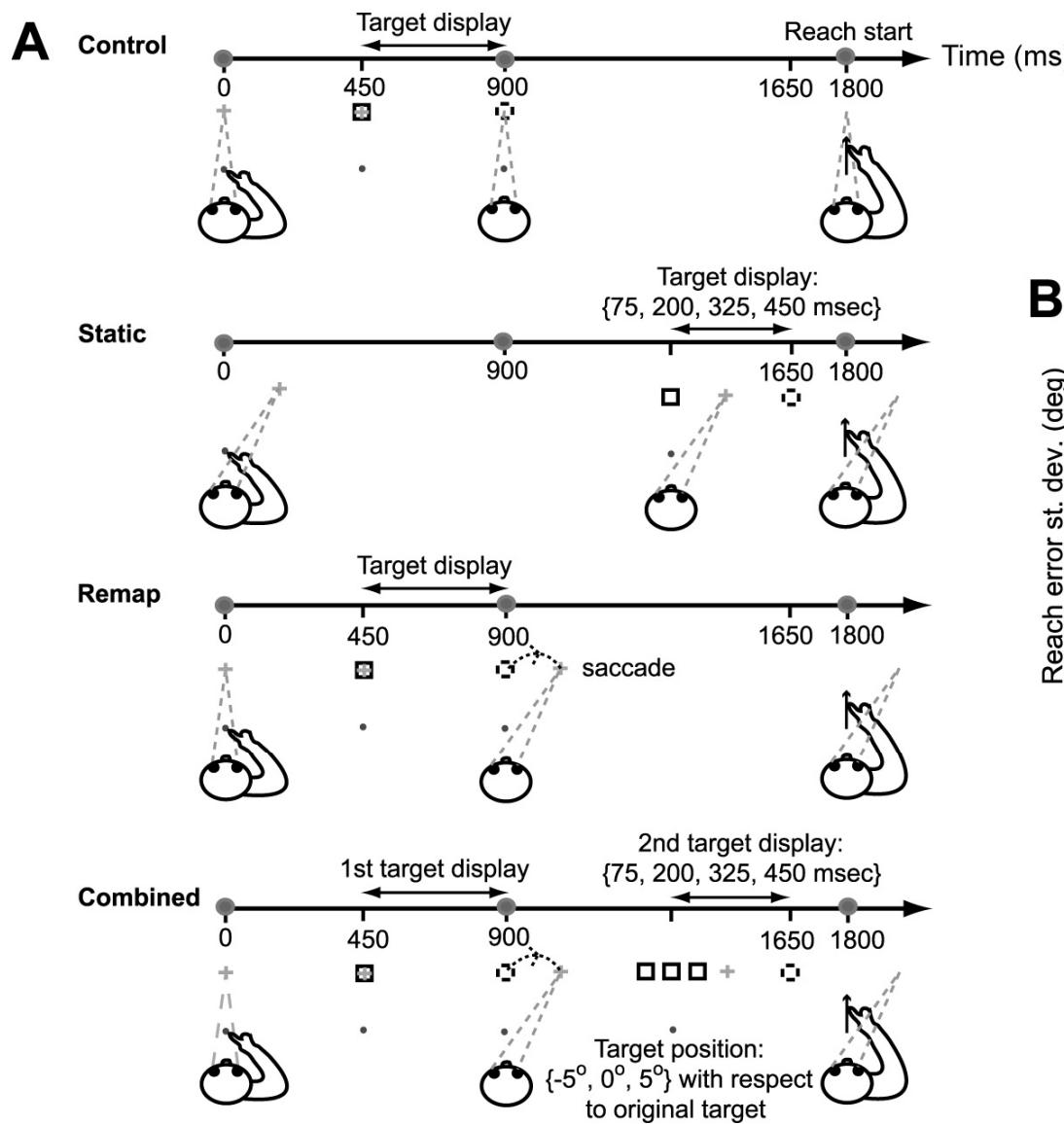
A

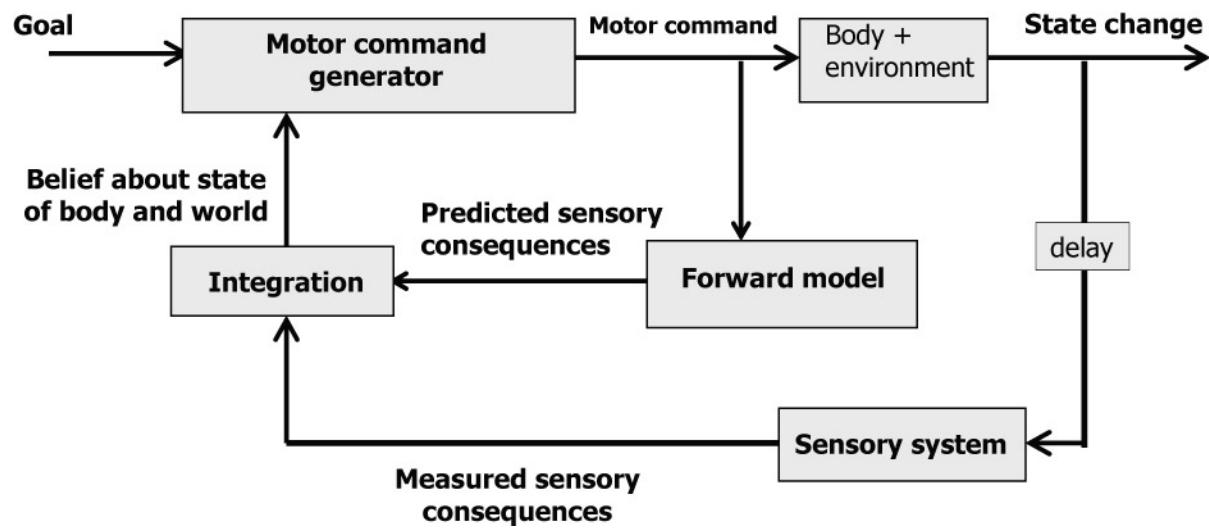


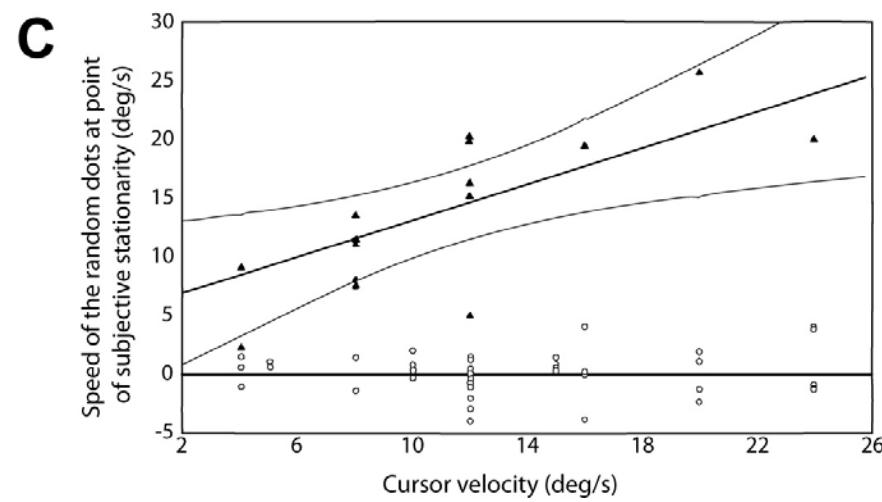
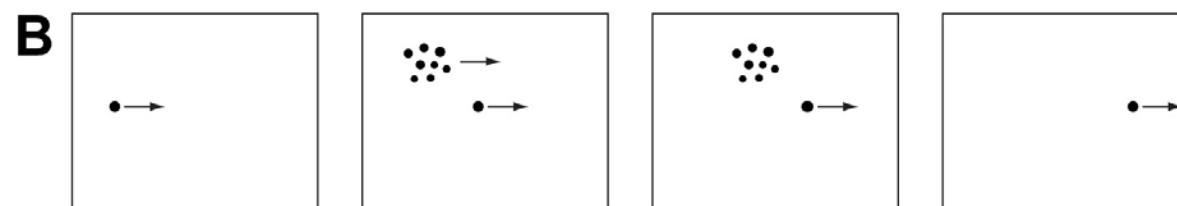
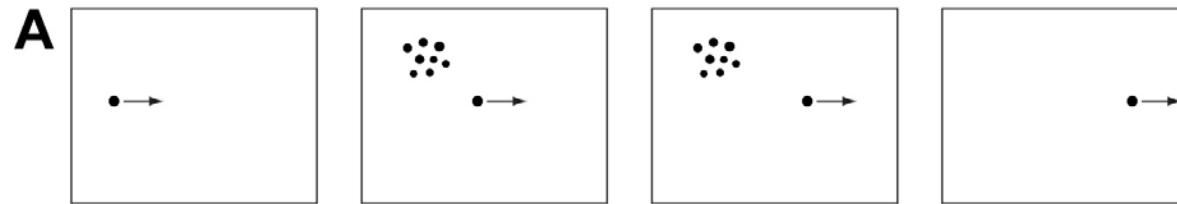
B

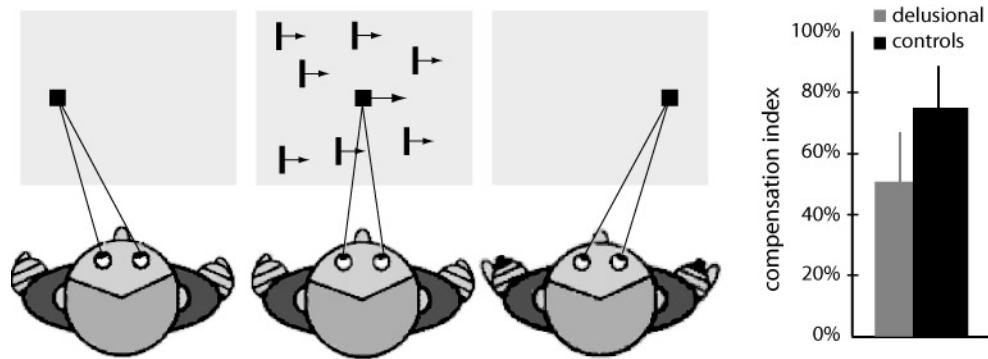


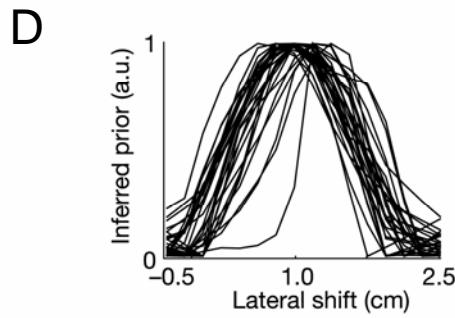
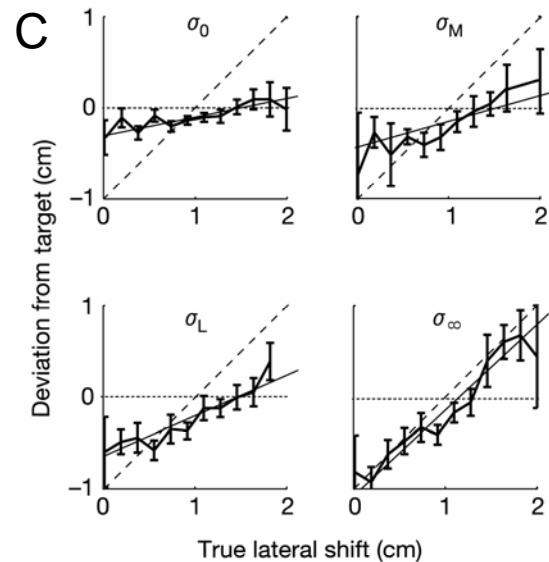
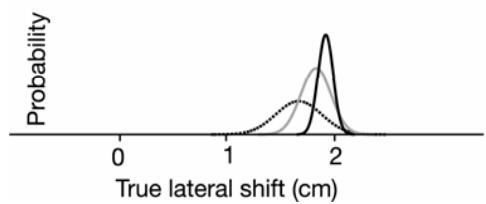
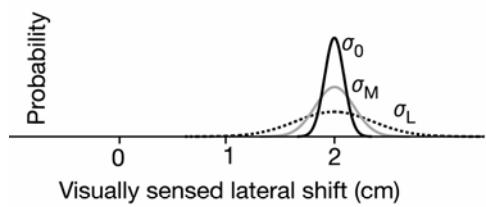
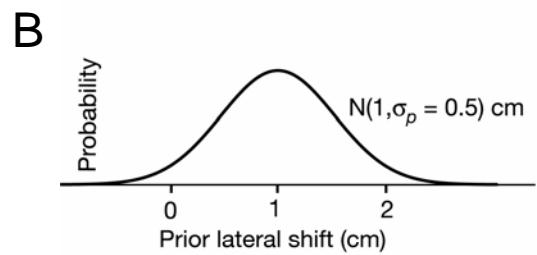
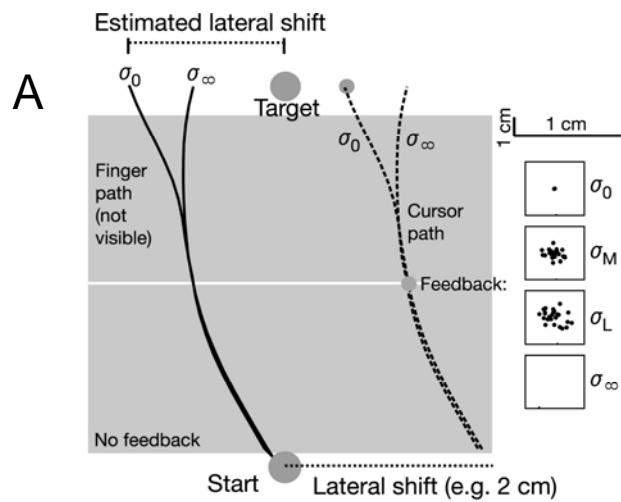


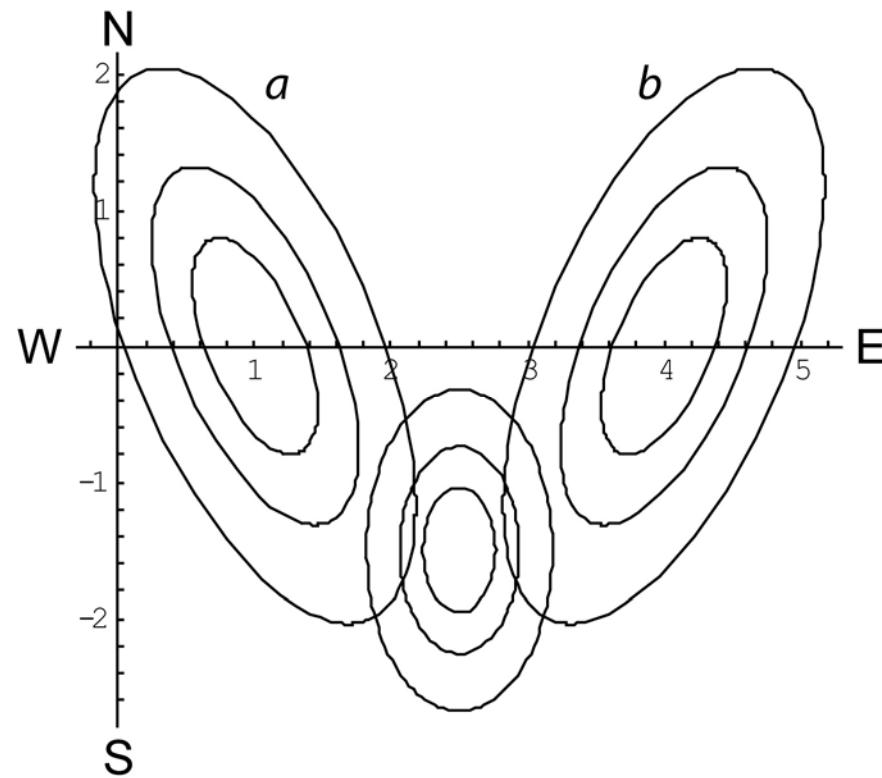


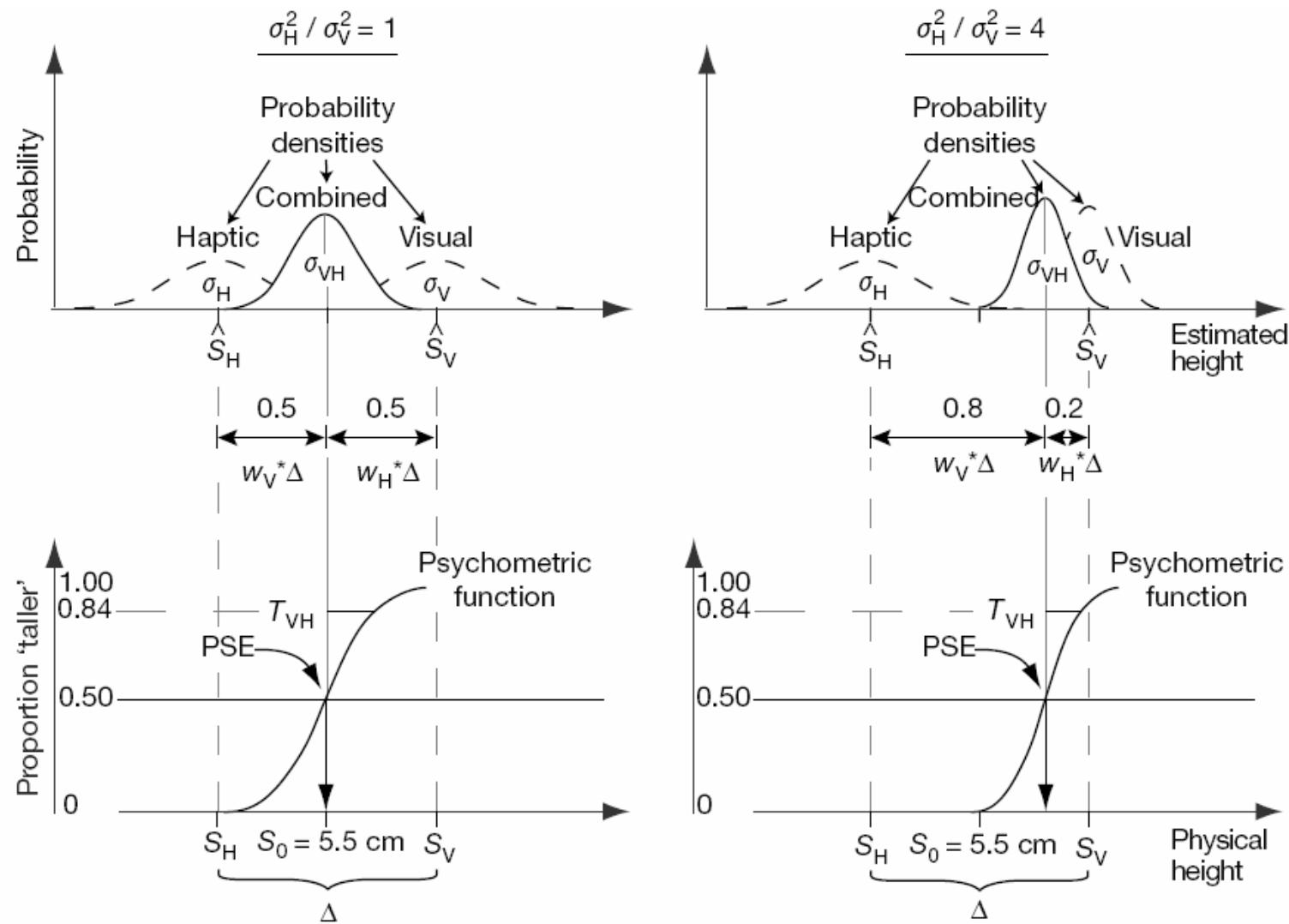


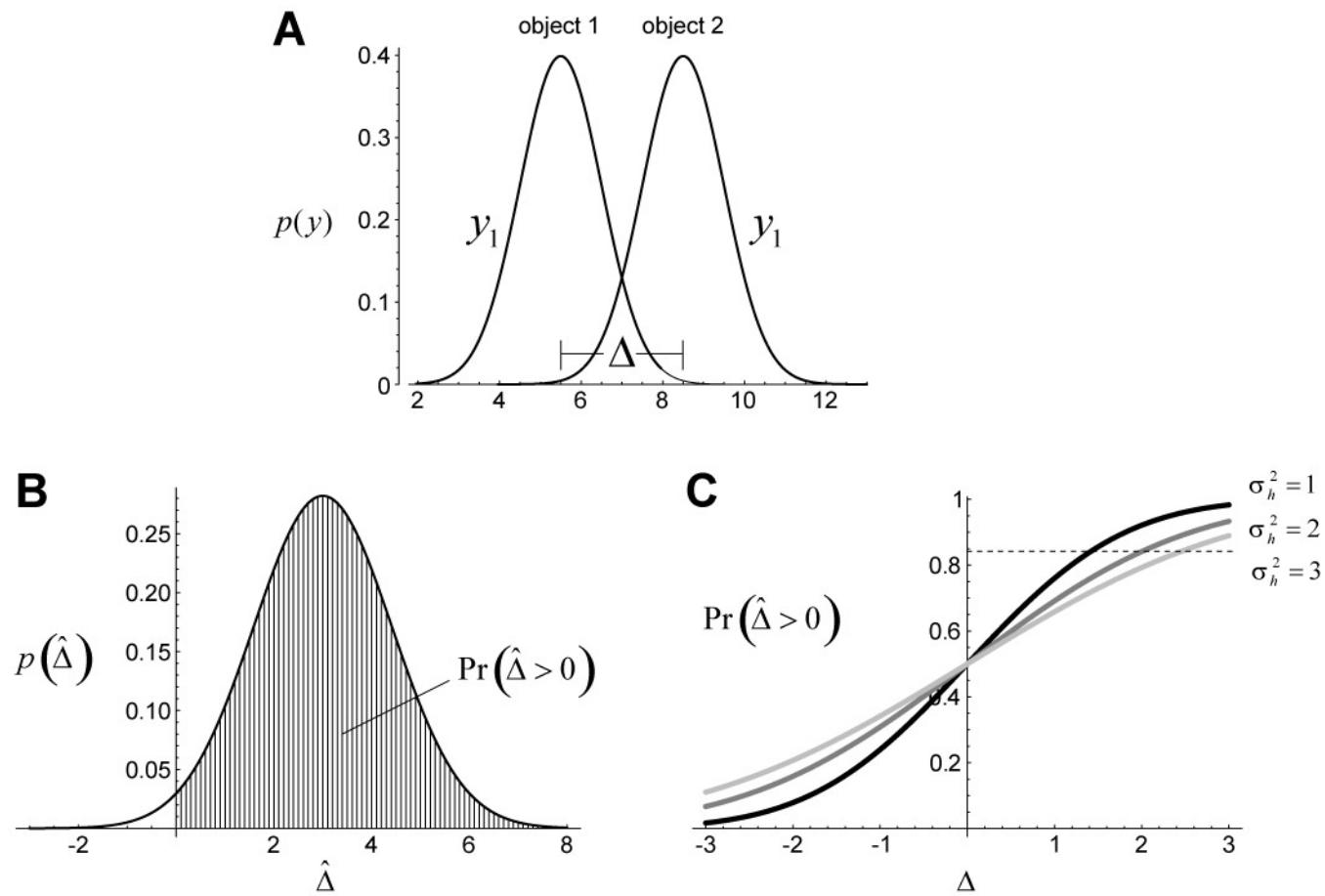


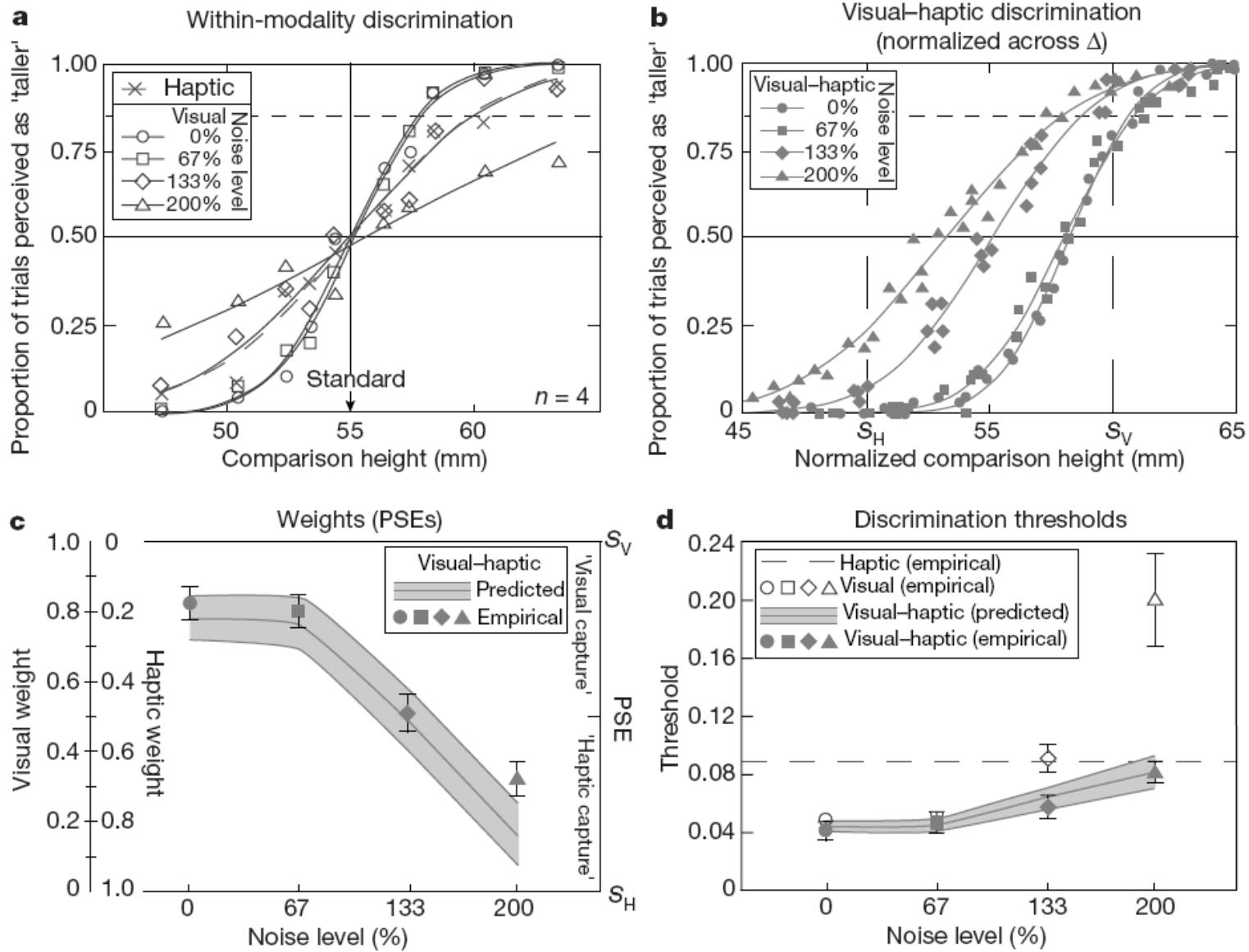




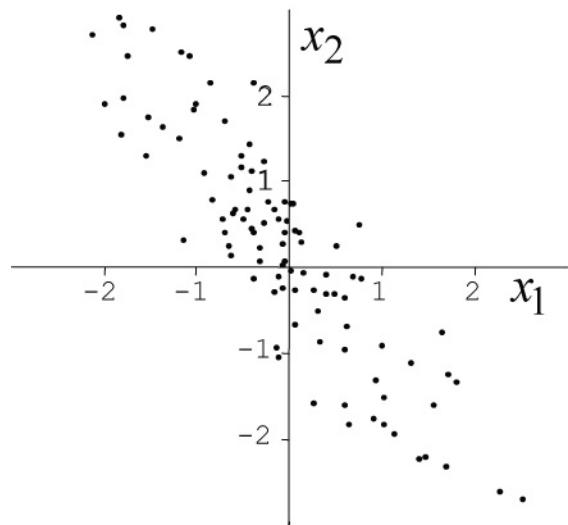




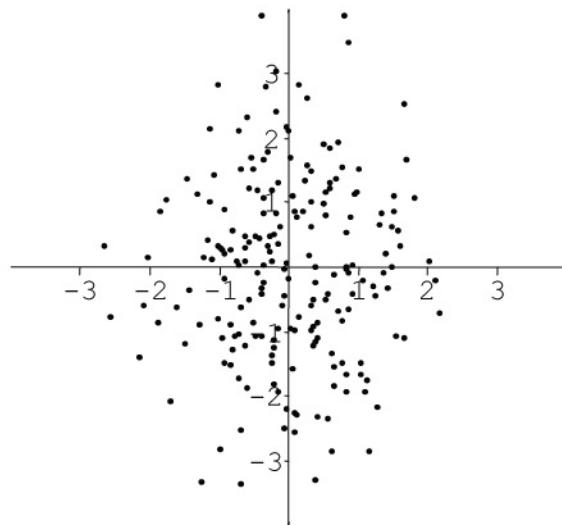




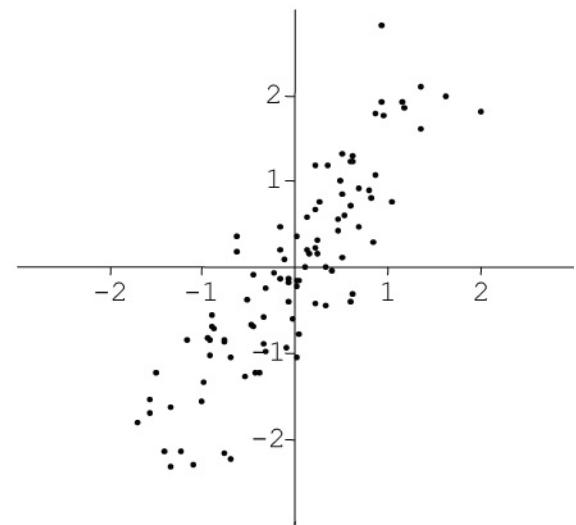
$$\mathbf{x} \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & -0.9\sqrt{2} \\ -0.9\sqrt{2} & 2 \end{bmatrix}\right)$$



$$\mathbf{x} \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0.1\sqrt{2} \\ 0.1\sqrt{2} & 2 \end{bmatrix}\right)$$



$$\mathbf{x} \sim N\left(\begin{bmatrix} 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 & 0.9\sqrt{2} \\ 0.9\sqrt{2} & 2 \end{bmatrix}\right)$$



**A**

$x_1$	$x_2$	$y^*$
1	0	0.5
1	0	0.5
1	0	0.5
1	0	0.5
0	1	0.5

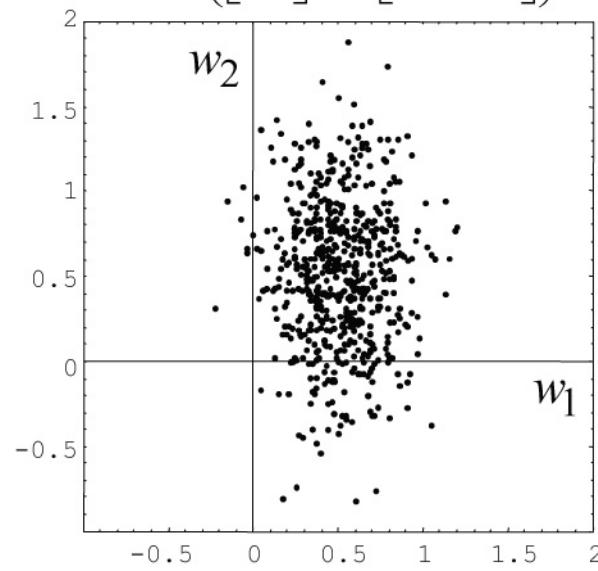
**B**

$x_1$	$x_2$	$y^*$
1	1	1
1	1	1
1	1	1
1	1	1
1	0	0.5

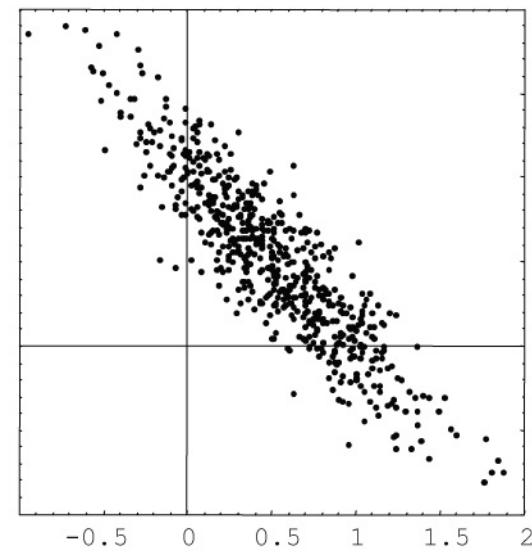
**C**

$x_1$	$x_2$	$y^*$
0	1	0.5
0	1	0.5
0	1	0.5
0	1	0.5
1	1	1

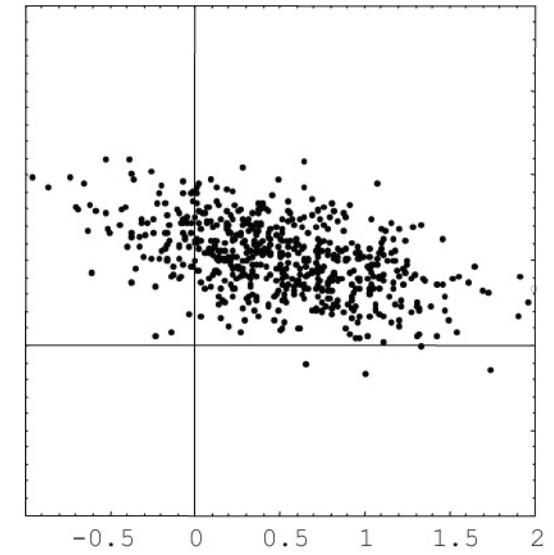
$$\mathbf{w} \sim N\left(\begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix}, \sigma^2 \begin{bmatrix} 0.25 & 0 \\ 0 & 1 \end{bmatrix}\right)$$

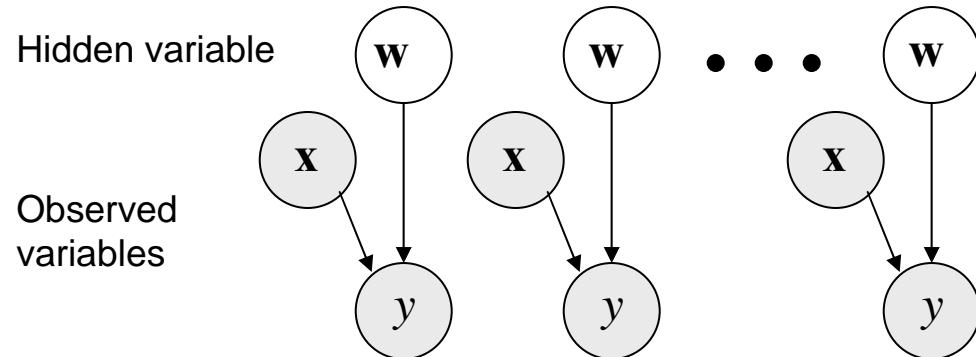


$$\mathbf{w} \sim N\left(\begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix}, \sigma^2 \begin{bmatrix} 1 & -1 \\ -1 & 1.25 \end{bmatrix}\right)$$



$$\mathbf{w} \sim N\left(\begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix}, \sigma^2 \begin{bmatrix} 1.25 & -0.25 \\ -0.25 & 0.25 \end{bmatrix}\right)$$





Generative model

$$\mathbf{w}^{(n+1)} = \mathbf{w}^{(n)}$$

$$y^{(n)} = \mathbf{x}^{(n)T} \mathbf{w} + \varepsilon^{(n)} \quad \varepsilon \sim N(0, \sigma^2)$$

Prior estimate of mean and variance of the hidden variable

$$\hat{\mathbf{w}}^{(1|0)}, P^{(1|0)}$$

$$\hat{\mathbf{w}}^{(n|n)} = \hat{\mathbf{w}}^{(n|n-1)} + \mathbf{k}^{(n)} \left( y^{(n)} - \mathbf{x}^{(n)T} \hat{\mathbf{w}}^{(n|n-1)} \right)$$

$$\mathbf{k}^{(n)} = \frac{P^{(n|n-1)} \mathbf{x}^{(n)}}{\mathbf{x}^{(n)T} P^{(n|n-1)} \mathbf{x}^{(n)} + \sigma^2}$$

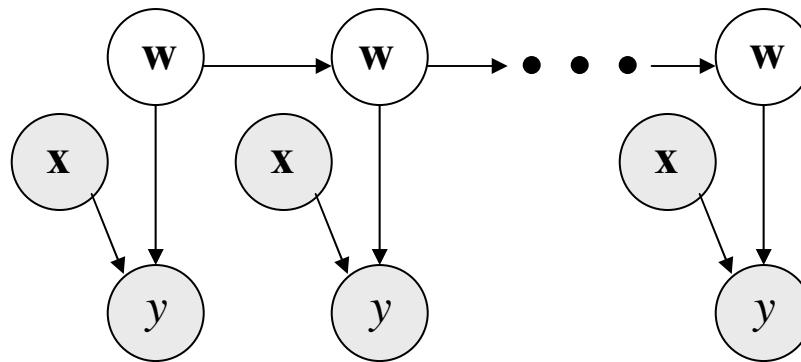
$$P^{(n|n)} = \left( I - \mathbf{k}^{(n)} \mathbf{x}^{(n)T} \right) P^{(n|n-1)}$$

Update of the estimate after making an observation

$$\hat{\mathbf{w}}^{(n+1|n)} = \hat{\mathbf{w}}^{(n|n)}$$

$$P^{(n+1|n)} = P^{(n|n)}$$

Forward projection of the estimate to the next trial



Generative model

$$\mathbf{w}^{(n+1)} = A\mathbf{w}^{(n)} + \boldsymbol{\varepsilon}_w^{(n)} \quad \boldsymbol{\varepsilon}_w \sim N(0, Q)$$

$$y^{(n)} = \mathbf{x}^{(n)T} \mathbf{w}^{(n)} + \boldsymbol{\varepsilon}_y^{(n)} \quad \boldsymbol{\varepsilon}_y \sim N(0, \sigma^2)$$

Prior estimate of mean and variance of the hidden variable

$$\left[ \hat{\mathbf{w}}^{(1|0)}, P^{(1|0)} \right]$$

Update of the estimate after making an observation

$$\begin{cases} \mathbf{k}^{(n)} = \frac{P^{(n|n-1)} \mathbf{x}^{(n)}}{\mathbf{x}^{(n)T} P^{(n|n-1)} \mathbf{x}^{(n)} + \sigma^2} \\ \hat{\mathbf{w}}^{(n|n)} = \hat{\mathbf{w}}^{(n|n-1)} + \mathbf{k}^{(n)} \left( y^{(n)} - \mathbf{x}^{(n)T} \mathbf{w}^{(n|n-1)} \right) \\ P^{(n|n)} = \left( I - \mathbf{k}^{(n)} \mathbf{x}^{(n)T} \right) P^{(n|n-1)} \end{cases}$$

Forward projection of the estimate to the next trial

$$\begin{cases} \hat{\mathbf{w}}^{(n+1|n)} = A\hat{\mathbf{w}}^{(n|n)} \\ P^{(n+1|n)} = AP^{(n|n)}A^T + Q \end{cases}$$

