

## Particle filters

→ Bayes filter

Histogram filter

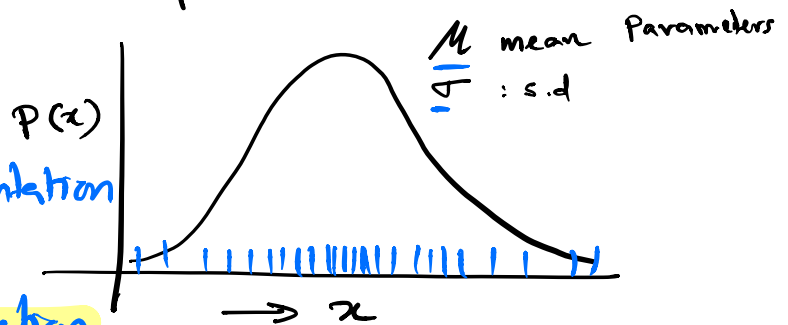
→ Alternative Nonparametric filter

→  $\text{bel}(x_t)$  represented by finite set of parameters

Overview :  $\text{bel}(x_t)$  represented by a set of random state samples

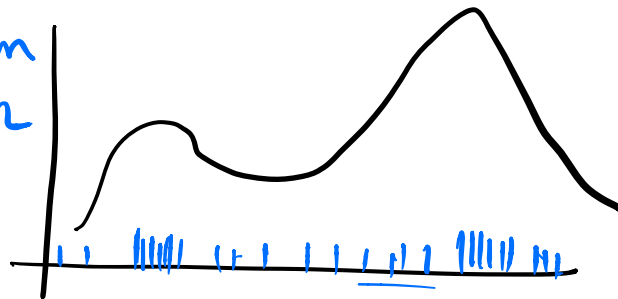
Example

non parametric representation



Nonlinear transformation

on the representation is going to be easier



## Particles

Samples of the distribution  $\text{bel}(x_t)$  are called particles

$M$  particles denoted as  $x_t^{[1]}, x_t^{[2]}, \dots, x_t^{[M]}$  together  $X_t$

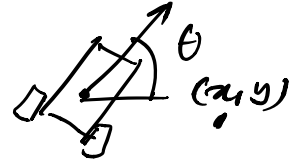
each  $x_t^{[m]}$   $1 \leq m \leq M$

$\in$  state vector for the system

E.g. Rearwheel vehicle model

$$\underline{x}_t^{[m]} = \langle \underline{x}_t^m, y_t^m, \theta_t^m \rangle$$

$[\underline{x}, y, \theta]$



$M \approx 1000$

$x_t^{[m]}$  should be included in  $\underline{x}_t$

with probability = bel( $x_t$ ) =  $P(x_t | z_{1:t}, u_{1:t})$

Relationship only hold asymptotically

$M \rightarrow \infty$

## Basic PF Algorithm

Recursive Algorithm

$$x_t = x_t^{[1]} \quad \dots \quad x_t^{[m]}$$

$X_t = x_t^{[1]}, x_t^{[2]}, \dots, x_t^{[M]}$  particles  
 Algorithm `Particle_filter`( $X_{t-1}, u_t, z_t$ ):  
 $\bar{X}_{t-1} = X_t = \emptyset$   
 for all  $m$  in  $[M]$  do:  
     sample  $x_t^{[m]} \sim p(x_t | u_t, x_{t-1}^{[m]})$   
      $w_t^{[m]} = p(z_t | x_t^{[m]})$   
      $\bar{X}_t = \bar{X}_t + \langle x_t^{[m]}, w_t^{[m]} \rangle$   
 end for  
 for all  $m$  in  $[M]$  do:  
     draw  $i$  with probability  $\propto w_t^{[i]}$   
     add  $x_t^{[i]}$  to  $X_t$   
 end for  
 return  $X_t$

Previous belief  
 Recent control int  $u_t$   
 measurement  $z_t$   
 $bel(x_{t-1})$   
 State transition model  
 measurement weights  
 measurement model  
 intermediate particles  
 Resampling step  
 importance sampling  
 "trick"

### Resampling in Particle filtering

"Survival of fittest"  
 We want to sample from  $bel(x_t)$   $X_t$   $f$   
 But we only get to sample from  $bel(x_t)$   $g$   
 E.g. compute prob  $x \in A$  acc density  
 function  $f$  but we only have samples  
 from  $g$ .

$$E_f [I(x \in A)] = \int f(x) I(x \in A) dx$$

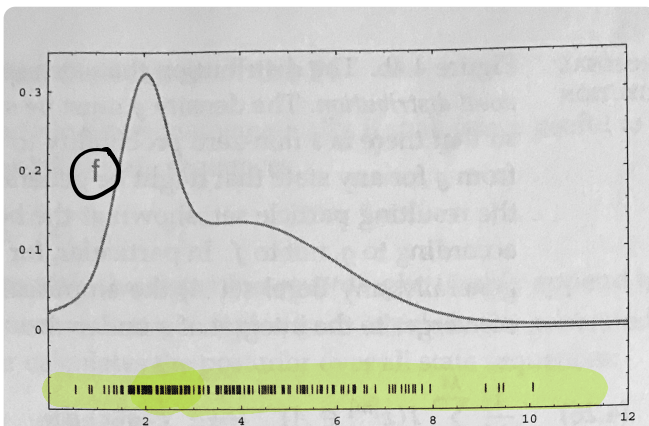
Provided  
 $f(x) > 0 \Rightarrow$   
 $g(x) > 0$

$$= \int \frac{f(x)}{g(x)} g(x) I(x \in A) dx$$

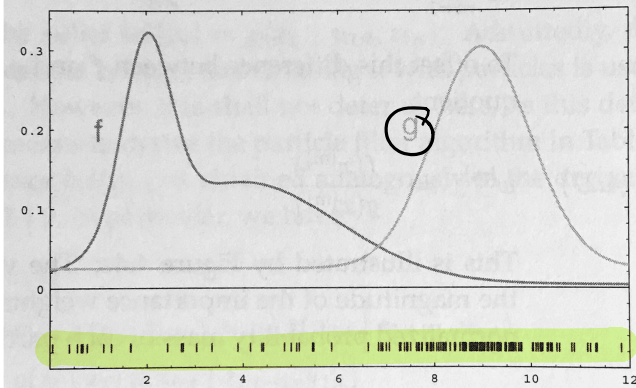
$$= w(x)$$

$$= \int w(x) g(x) I(x \in A) dx$$

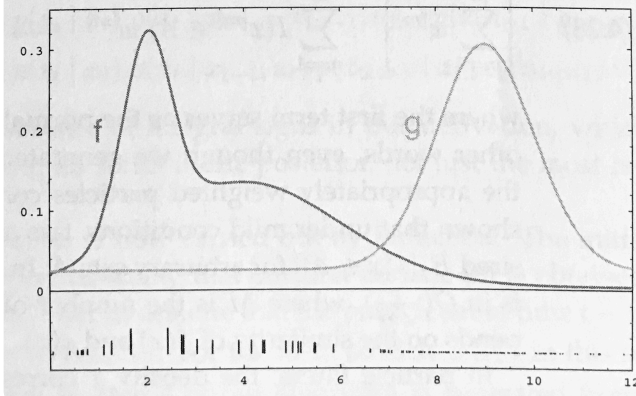
$$= E_g [w(x) I(x \in A)]$$



bel( $x_t$ )



Samples from  $g$   
 bel( $x_t$ )



Samples  $f$   
 obtained  
 by weighting  
 $\frac{f(x)}{g(x)}$  to  
 each sample  
 of  $g$

2,