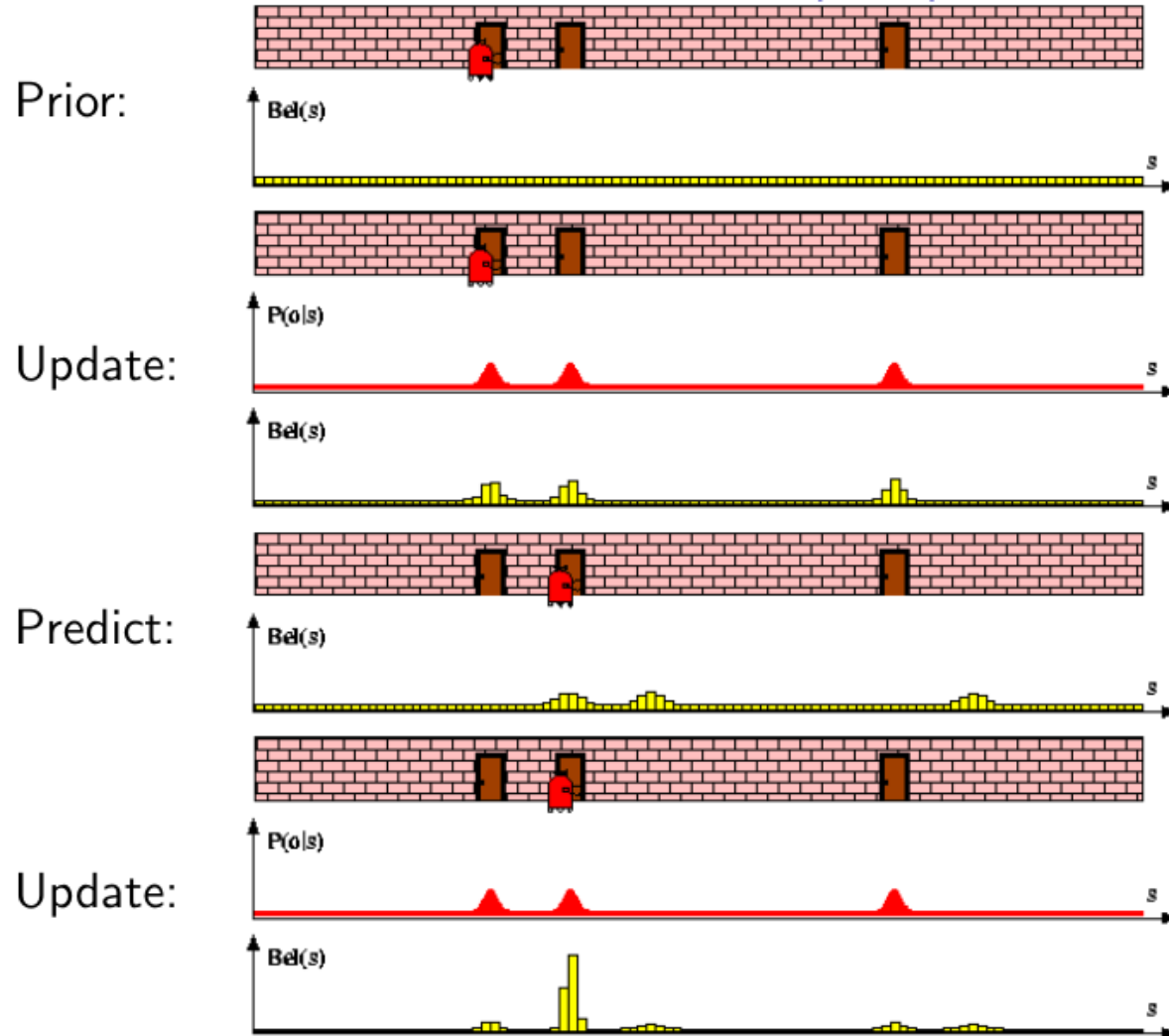


Bayes filter on 1D localization (Homework)

Bayes filter on 1D localization



PR book Figure 8.1

You can implement a discrete bayes filter for the 1D robot localization.

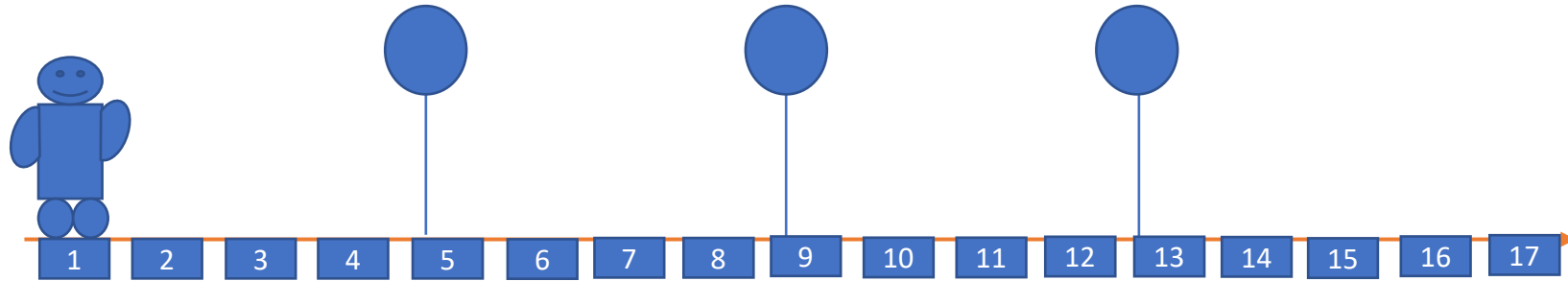
```

1:  Algorithm Discrete_Bayes_filter( $\{p_{k,t-1}\}, u_t, z_t$ ):
2:    for all  $k$  do
3:       $\bar{p}_{k,t} = \sum_i p(X_t = x_k \mid u_t, X_{t-1} = x_i) p_{i,t-1}$ 
4:       $p_{k,t} = \eta p(z_t \mid X_t = x_k) \bar{p}_{k,t}$ 
5:    endfor
6:    return  $\{p_{k,t}\}$ 
    
```

Table 4.1 The discrete Bayes filter. Here x_i, x_k denote individual states.

PR book pp. 87

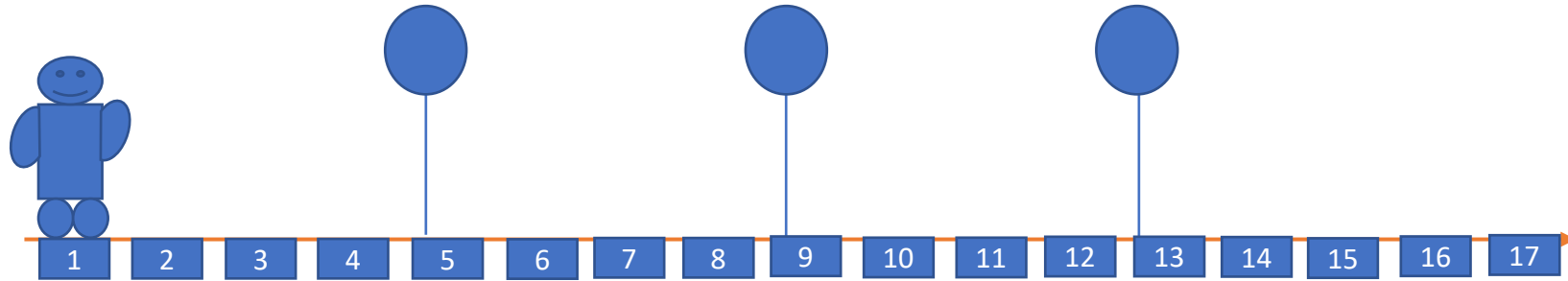
Bayes filter on simplified 1D localization (Homework)



We further simplify the scenario as follows:

- world: discrete grids denoted as spots #1 ~ #40
- control command: move forward one grid
- landmarks: three poles with known locations
- sensor observations: pole detected/not detected in front of the robot

Bayes filter on simplified 1D localization (Homework)



Notations:

$P(L_i)$ = The probability the robot is located in location i

$P(D)$ = The probability that a pole is detected

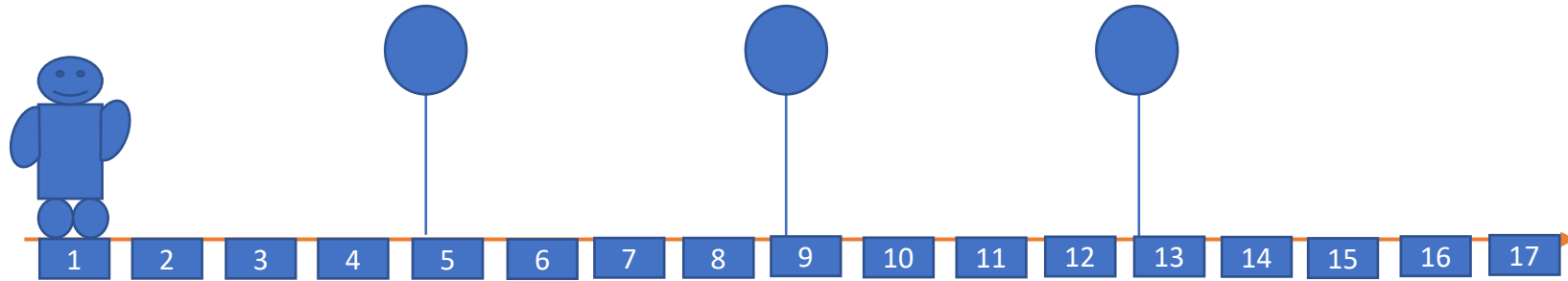
$P(!D)$ = The probability that a pole is not detected

$P(D \mid L_i)$ = The probability of a pole being detected, given the robot is located at the location i

$P(L_i \mid D) = ?$

$P(L_i \mid !D) = ?$

Bayes filter on simplified 1D localization (Homework)

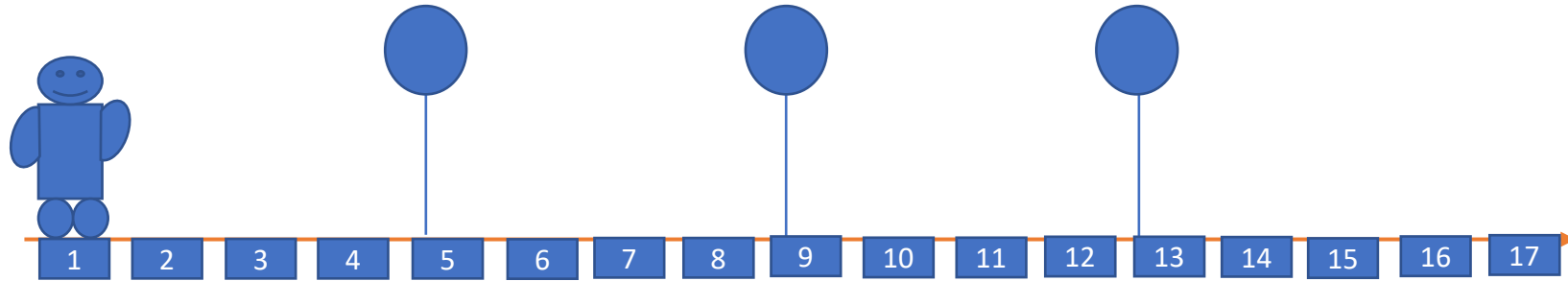


What happens when the robot moves?

1. Shift the $P(L_i)$ think: what is $P(L_i)$ now? e.g., $P(L_5) \leftarrow P(L_4 | D)$
2. Do Bayes rule for all $P(L_i | D)$ again

Complete the assignment 2-1

Bayes filter on simplified 1D localization (Homework)



What if robot moves with uncertainties:

- sometimes, the control command asks the robot to move one unit ahead, but the robot can accidentally move two units ahead with the probability 10%.

Complete the assignment 2-2