

Trabalho #6

Simular os algoritmos apresentados no capítulo 8 do livro [THRUN ETAL:2006] .

Tabelas com os algoritmos:

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1: Algorithm Grid_localization( $\{p_{k,t-1}\}, u_t, z_t, m$ ):
2:   for all  $k$  do
3:      $\bar{p}_{k,t} = \sum_i p_{i,t-1} \text{motion\_model}(\text{mean}(\mathbf{x}_k), u_t, \text{mean}(\mathbf{x}_i))$ 
4:      $p_{k,t} = \eta \bar{p}_{k,t} \text{measurement\_model}(z_t, \text{mean}(\mathbf{x}_k), m)$ 
5:   endfor
6:   return  $\{p_{k,t}\}$ 

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1: Algorithm MCL( $\mathcal{X}_{t-1}, u_t, z_t, m$ ):
2:    $\bar{\mathcal{X}}_t = \mathcal{X}_t = \emptyset$ 
3:   for  $m = 1$  to  $M$  do
4:      $x_t^{[m]} = \text{sample\_motion\_model}(u_t, x_{t-1}^{[m]})$ 
5:      $w_t^{[m]} = \text{measurement\_model}(z_t, x_t^{[m]}, m)$ 
6:      $\bar{\mathcal{X}}_t = \bar{\mathcal{X}}_t + \langle x_t^{[m]}, w_t^{[m]} \rangle$ 
7:   endfor
8:   for  $m = 1$  to  $M$  do
9:     draw  $i$  with probability  $\propto w_t^{[i]}$ 
10:    add  $x_t^{[i]}$  to  $\mathcal{X}_t$ 
11:   endfor
12:   return  $\mathcal{X}_t$ 

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1: Algorithm Augmented_MCL( $\mathcal{X}_{t-1}, u_t, z_t, m$ ):
2:   static  $w_{\text{slow}}, w_{\text{fast}}$ 
3:    $\bar{\mathcal{X}}_t = \mathcal{X}_t = \emptyset$ 
4:   for  $m = 1$  to  $M$  do
5:      $x_t^{[m]} = \text{sample\_motion\_model}(u_t, x_{t-1}^{[m]})$ 
6:      $w_t^{[m]} = \text{measurement\_model}(z_t, x_t^{[m]}, m)$ 
7:      $\bar{\mathcal{X}}_t = \bar{\mathcal{X}}_t + \langle x_t^{[m]}, w_t^{[m]} \rangle$ 
8:      $w_{\text{avg}} = w_{\text{avg}} + \frac{1}{M} w_t^{[m]}$ 
9:   endfor
10:   $w_{\text{slow}} = w_{\text{slow}} + \alpha_{\text{slow}}(w_{\text{avg}} - w_{\text{slow}})$ 
11:   $w_{\text{fast}} = w_{\text{fast}} + \alpha_{\text{fast}}(w_{\text{avg}} - w_{\text{fast}})$ 
12:  for  $m = 1$  to  $M$  do
13:    with probability  $\max\{0.0, 1.0 - w_{\text{fast}}/w_{\text{slow}}\}$  do
14:      add random pose to  $\mathcal{X}_t$ 
15:    else
16:      draw  $i \in \{1, \dots, N\}$  with probability  $\propto w_t^{[i]}$ 
17:      add  $x_t^{[i]}$  to  $\mathcal{X}_t$ 
18:    endwith
19:  endfor
20:  return  $\mathcal{X}_t$ 

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1: Algorithm KLD_Sampling_MCL( $\mathcal{X}_{t-1}, u_t, z_t, m, \varepsilon, \delta$ ):
2:    $\mathcal{X}_t = \emptyset$ 
3:    $M = 0, M_\chi = 0, k = 0$ 
4:   for all  $b$  in  $H$  do
5:      $b = \text{empty}$ 
6:   endfor
7:   do
8:     draw  $i$  with probability  $\propto w_{t-1}^{[i]}$ 
9:      $x_t^{[M]} = \text{sample\_motion\_model}(u_t, x_{t-1}^{[i]})$ 
10:     $w_t^{[M]} = \text{measurement\_model}(z_t, x_t^{[M]}, m)$ 
11:     $\mathcal{X}_t = \mathcal{X}_t + \langle x_t^{[M]}, w_t^{[M]} \rangle$ 

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12:    if  $x_t^{[M]}$  falls into empty bin  $b$  then
13:       $k = k + 1$ 
14:       $b = \text{non-empty}$ 
15:      if  $k > 1$  then
16:         $M_\chi := \frac{k-1}{2\varepsilon} \left\{ 1 - \frac{2}{9(k-1)} + \sqrt{\frac{2}{9(k-1)}} z_{1-\delta} \right\}^3$ 
17:      endif
18:       $M = M + 1$ 
19:      while  $M < M_\chi$  or  $M < M_{\chi_{\min}}$ 
20:        return  $\mathcal{X}_t$ 

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1: Algorithm test_range_measurement( $z_t^k, \bar{\mathcal{X}}_t, m$ ):
2:    $p = q = 0$ 
3:   for  $m = 1$  to  $M$  do
4:      $p = p + z_{\text{short}} \cdot p_{\text{short}}(z_t^k | x_t^{[m]}, m)$ 
5:      $q = q + z_{\text{hit}} \cdot p_{\text{hit}}(z_t^k | x_t^{[m]}, m) + z_{\text{short}} \cdot p_{\text{short}}(z_t^k | x_t^{[m]}, m)$ 
6:        $+ z_{\text{max}} \cdot p_{\text{max}}(z_t^k | x_t^{[m]}, m) + z_{\text{rand}} \cdot p_{\text{rand}}(z_t^k | x_t^{[m]}, m)$ 
7:   endfor
8:   if  $p/q \leq \chi$  then
9:     return accept
10:  else
11:    return reject
12:  endif

```

Obs.: Pesquisar detalhes do Algoritmo KLD Sampling MCL no artigo

- [1] DIETER FOX
Adapting the Sample Size in Particle Filters Through KLD-Sampling.
 The International Journal of Robotics Research, Vol. 22, No. 12, pp. 985-1003, **2003**.
[\[doi: 10.1177/0278364903022012001\]](https://doi.org/10.1177/0278364903022012001)

Material disponível no Moodle

- Notas de aula.

Referências

- [2] SEBASTIAN THRUN, WOLFRAM BURGARD & DIETER FOX
Probabilistic robotics.
 MIT Press, **2006**.
 Link: <http://probabilistic-robotics.informatik.uni-freiburg.de/>
- [3] HOWIE CHOSSET, KEVIN LYNCH, SETH HUTCHINSON, GEORGE KANTOR, WOLFRAM BURGARD, LYDIA KAVRAKI & SEBASTIAN THRUN
Principles of Robot Motion. Theory, Algorithms, and Implementations.
 MIT Press, **2005**.
 Link: <http://biorobotics.ri.cmu.edu/book/>

Contém uma descrição detalhada do filtro de Kalman e do EKF.

- [4] GREGOR KLANČAR, ANDREJ ZDEŠAR, SAŠO BLAŽIC & IGOR ŠKRJANC
Wheeled Mobile Robotics. From Fundamentals Towards Autonomous Systems.
Butterworth-Heinemann, 2017.
Link: <http://booksite.elsevier.com/9780128042045/manuscript.php>

Contém códigos em Matlab.

Apresentações

- Os grupos terão cerca de 25 minutos para fazer as apresentações.
- As apresentações serão realizadas na seguinte data:



Avaliação do trabalho

Preparar e enviar por email:

1. Relatório contendo a descrição dos algoritmos, resultados das simulações e discussão dos resultados.
2. Códigos dos programas utilizados nas simulações.
3. Slides preparados para a apresentação do trabalho.

Grupos

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