Trabalho #4

Simular os algoritmos apresentados no capítulo 6 do livro [Thrun etal: 2006].

Tabelas com os algoritmos:

```
 \begin{array}{lll} 1: & \textbf{Algorithm beam\_range\_finder\_model}(z_t, x_t, m) \text{:} \\ 2: & q = 1 \\ 3: & \textit{for } k = 1 \textit{ to } K \textit{ do} \\ 4: & \textit{compute } z_t^{k*} \textit{ for the measurement } z_t^k \textit{ using ray casting} \\ 5: & p = z_{\text{hit}} \cdot p_{\text{hit}}(z_t^k \mid x_t, m) + z_{\text{short}} \cdot p_{\text{short}}(z_t^k \mid x_t, m) \\ 6: & +z_{\text{max}} \cdot p_{\text{max}}(z_t^k \mid x_t, m) + z_{\text{rand}} \cdot p_{\text{rand}}(z_t^k \mid x_t, m) \\ 7: & q = q \cdot p \\ 8: & \textit{return } q \\ \end{array}
```

```
1:
               Algorithm learn_intrinsic_parameters (Z, X, m):
2:
                      repeat until convergence criterion satisfied
3:
                             for all z_i in Z do
4:
                                    \eta = [p_{\text{hit}}(z_i \mid x_i, m) + p_{\text{short}}(z_i \mid x_i, m)]
                                         + \, p_{\max}(z_i \mid x_i, m) + p_{\mathrm{rand}}(z_i \mid x_i, m) \, \big]^{-1}
5:
                                    calculate z^*
6:
                                    e_{i,\mathrm{hit}} = \eta \ p_{\mathrm{hit}}(z_i \mid x_i, m)
7:
                                    e_{i,\text{short}} = \eta \ p_{\text{short}}(z_i \mid x_i, m)
8:
                                    e_{i,\max} = \eta \; p_{\max}(z_i \mid x_i, m)
9:
                                    e_{i,\text{rand}} = \eta \; p_{\text{rand}}(z_i \mid x_i, m)
                             z_{\text{hit}} = |Z|^{-1} \sum_{i} e_{i,\text{hit}}
10:
                             z_{\rm short} = |Z|^{-1} \sum_{i} e_{i, \rm short}
11:
                             z_{\text{max}} = |Z|^{-1} \sum_{i} e_{i,\text{max}}
12:
                             z_{\rm rand} = |Z|^{-1} \sum_{i} e_{i, {\rm rand}}
13:
14:
                             \sigma_{\text{hit}} = \sqrt{\frac{1}{\sum_{i} e_{i,\text{hit}}} \sum_{i} e_{i,\text{hit}} (z_i - z_i^*)^2}
15:
16:
                      \textit{return}\ \Theta = \{z_{\text{hit}}, z_{\text{short}}, z_{\text{max}}, z_{\text{rand}}, \sigma_{\text{hit}}, \lambda_{\text{short}}\}
```

```
1:
           {\bf Algorithm\ likelihood\_field\_range\_finder\_model(} z_t, x_t, m) {\bf :}
2:
                   a = 1
 3:
                   for all k do
 4:
                         if z_t^k \neq z_{\max}
 5:
                                x_{z_t^k} = x + x_{k,\text{sens}} \cos \theta - y_{k,\text{sens}} \sin \theta + z_t^k \cos(\theta + \theta_{k,\text{sens}})
                                y_{z_t^k} = y + y_{k,\text{sens}} \cos \theta + x_{k,\text{sens}} \sin \theta + z_t^k \sin(\theta + \theta_{k,\text{sens}})
 6:
 7:
                                dist = \min_{x',y'} \left\{ \sqrt{(x_{z_t^k} - x')^2 + (y_{z_t^k} - y')^2} \middle| \langle x', y' \rangle \text{ occupied in } m \right\}
 8:
                                q = q \cdot \left(z_{\text{hit}} \cdot \mathbf{prob}(dist, \sigma_{\text{hit}}) + \frac{z_{\text{random}}}{z_{\text{max}}}\right)
 9:
                   return q
```

```
 \begin{array}{ll} 1: & \textbf{Algorithm landmark\_model\_known\_correspondence}(f_t^i, c_t^i, x_t, m) : \\ 2: & j = c_t^i \\ 3: & \hat{r} = \sqrt{(m_{j,x} - x)^2 + (m_{j,y} - y)^2} \\ 4: & \hat{\phi} = \operatorname{atan2}(m_{j,y} - y, m_{j,x} - x) \\ 5: & q = \operatorname{prob}(r_t^i - \hat{r}, \sigma_r) \cdot \operatorname{prob}(\phi_t^i - \hat{\phi}, \sigma_\phi) \cdot \operatorname{prob}(s_t^i - s_j, \sigma_s) \\ 6: & \operatorname{return} q \\ \end{array}
```

```
1: Algorithm sample_landmark_model_known_correspondence(f_t^i, c_t^i, m):

2: j = c_t^i
3: \hat{\gamma} = \operatorname{rand}(0, 2\pi)
4: \hat{r} = r_t^i + \operatorname{sample}(\sigma_r)
5: \hat{\phi} = \phi_t^i + \operatorname{sample}(\sigma_{\phi})
6: x = m_{j,x} + \hat{r} \cos \hat{\gamma}
7: y = m_{j,y} + \hat{r} \sin \hat{\gamma}
8: \theta = \hat{\gamma} - \pi - \hat{\phi}
9: \operatorname{return}(x \ y \ \theta)^T
```

Material disponível no Moodle

• Notas de aula.

Referências

[1] Sebastian Thrun, Wolfram Burgard & Dieter Fox

Probabilistic robotics.

MIT Press, 2006.

Link: http://probabilistic-robotics.
informatik.uni-freiburg.de/

[2] HOWIE CHOSET, KEVIN LYNCH, SETH HUTCHIN-SON, 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.

[3] Gregor Klancar, 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 20 minutos para fazer as apresentações.
- As apresentações serão realizadas na seguinte data:

16/dez (4a. feira)

Avaliação do trabalho

Preparar e enviar por email:

- 1. Relatório contendo a descrição dos algoritmos, resultados das simulações e discussão dos resultados.
- $2.\ {\rm C\'odigos}$ dos programas utilizados nas simulações.
- $3.\,$ Slides preparados para a apresentação do trabalho.

Grupos

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