Tom Krajník	Autonomous Navigation	AIC@CTU	Tom Krajník	Autonomous Navigation	AIC@CTU	Tom Krajník	Autonomous Navigation	AIC@CTU									
				Navigation			Navigation										
Artificial intelligence in robotics 2019 Simultanneous localisation and mapping Tom Krajník FEL ČVUT			 The art of getting from one place to another, safely and efficiently. The process of monitoring and controlling the movement of a craft or vehicle from one place to another. The activity of accurately ascertaining one's position and planning and following a route. 			 The art of getting from one place to another, safely and efficiently. The process of monitoring and controlling the movement of a craft or vehicle from one place to another. The activity of accurately ascertaining one's position and planning and following a route. 											
										Nov 2019					C	A C	in succession of the second
									om Krajník	Artificial intelligence in robotics	1/7	"Where am	I?", "Where am I going?", "How do Artificial intelligence in robotics		Localisation,	Mapping, Artificial intelligence in robotics	Motion planning
om Krajník	Autonomous Navigation	AIC@CTU	Tom Krajník	Autonomous Navigation	AIC@CTU	Tom Krajník	Autonomous Navigation	AIC@CTU									
	Lecture intro			What to remember			What to remember										
Autonomous navigation in mobile robotics: 1. Map-less navigation			Overview • (observations, map, position) \leftrightarrow (map, position)			Probabilistic formulation of full SLAM											
 observations translate to motion commands unknown, structured (roads, corridors, lanes) environments observations → commands Map-based navigation observations and map data translate to motion commands known (mapped), (un)structured environments (observations, map) → commands Map-building-based navigation observations and map data translate to both commands and map (un)structured environments 			 essential component of navigation systems but it does not solve navigation by itself. Odometry vs. SLAM the drift issue loop closure is the difference! 			$p(x_{0:T}, m o_{1:T}, u_{1:T}).$ (1)											
						Probabilistic formulation of 'online' SLAM											
						$p(x_T, m o_{1:T}, u_{1:T}).$ (2) • $x_{0:T}$ - trajectory • m - map • $o_{1:T}$ - sequence of observations											
									• (obse	rvations) \rightarrow (commands, map)					• $u_{1:T}$ - seque	nce of commands	
									m Krajník m Krajník	Artificial intelligence in robotics Autonomous Navigation	3 / 7 AIC@CTU	Tom Krajník Tom Krajník	Artificial intelligence in robotics Autonomous Navigation	4 / 7 AIC@CTU	Tom Krajník	Artificial intelligence in robotics	
										What to remember			Further study				
Map types • topological, • dense metri	Uncertainty models • (extended) Ka		https://	dy iss: Introduction to Robot Mapping www.youtube.com/watch?v=wVsfCnyt5 a et al.: Past, Present, and Future of Sir													

- sparse metric,
- hybrid.

- mixture models,
- particle filter.

- Localization and Mapping: Toward the Robust-Perception Age. IEEE T-RO 2018.
- Grissetti, Stachniss et al: Tutorial on Graph-Based SLAM. ITS Magazine