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Abbreviations

MR Mobile robot

IR Infrared

LIDAR Light detection and ranging

TLD Tracking learning detection

KLT Kanade lucas tomasi

KF Kalman filter

GPS Global positioning system

RFID Radio frequency identification

SVM Support vector machines

HOG Histogram of oriented gradients

ROS Robot operating system

K Kalman gain

V-J Viola jones

COG Center of gravity

ICC Instantaneous center of curvature

2D Two dimensional

3D Three dimensional

rhs Right hand side

fps Frame per second

Symbols

ii(x, y)	Image integral
(V_L,V_R)	Linear velocity
$(\omega_{\!\scriptscriptstyle L},\omega_{\!\scriptscriptstyle R})$	Angular velocity
V_l	Left wheel velocity
V_r	Right wheel velocity
L	Distance between right wheel and left wheel (axle length)
r	Radius of the wheel
$I_{\text{gray}}(x, y)$	Pixel values in grayscale format
$W_1, W_2 W_3$	Weights
C_n	Weighted cost function
n	Number of handleable objects
A_n	Normalized object area
$O(b^d)$	B is the branching factor and d is the depth of the goal node.
d_{tg}	Normalized distance from target (object location) to goal
d_{st}	Normalized distance from start to the target value
\hat{x}_k	New updated estimate
\hat{x}_k^-	Best estimate prior to receiving a measurement at time t_k