



Data-driven Conflict Detection Enhancement in Closest Point of Approach Problem

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Data-driven Conflict Detection Enhancement with Machine Learning

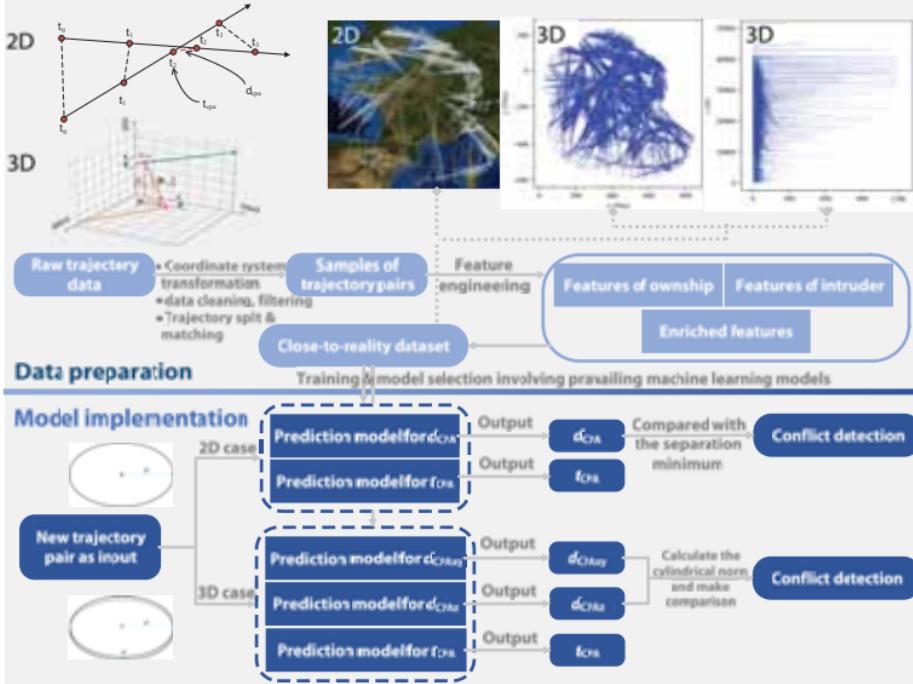
Objective:

To propose a novel data-driven conflict detection framework with machine learning for performance enhancement in actual operations.

Methodology:

Closest Point of Approach (CPA): Positions at which two dynamically moving objects reach their closest possible distance. It is a key concept in the algorithmic level for conflict detection.

Main problem of conventional model: Assumption cannot be ensured in real operations.



Results:

	CPA prediction						Conflict detection								
	2D case:			3D case:			2D case:			3D case:					
Models	$d_{CPA, \text{min}}$ MAE RMSE	$t_{CPA, \text{min}}$ MAE RMSE	Models	$d_{CPA, \text{min}}$ MAE RMSE	$t_{CPA, \text{min}}$ MAE RMSE	Models	$d_{CPA, \text{min}}$ MAE RMSE	$t_{CPA, \text{min}}$ MAE RMSE	Models	$d_{CPA, \text{min}}$ MAE RMSE	$t_{CPA, \text{min}}$ MAE RMSE				
Baseline	0.76	8.09	93.05	766.31											
RF	22.23	34.36	237.11	288.63											
FFNNs	0.29	10.29	98.64	149.12											
FFNNs ₀	1.76	3.31	10.10	33.73											
KNN	3.79	10.07	36.81	127.11											
GBM	3.62	2.03	28.01	65.66											
GBM ₀	2.14	4.20	34.26	72.24											
Models	$d_{CPA, \text{min}}$ MAE RMSE	$t_{CPA, \text{min}}$ MAE RMSE	Models	$d_{CPA, \text{min}}$ MAE RMSE	$t_{CPA, \text{min}}$ MAE RMSE	Models	$d_{CPA, \text{min}}$ MAE RMSE	$t_{CPA, \text{min}}$ MAE RMSE	Models	$d_{CPA, \text{min}}$ MAE RMSE	$t_{CPA, \text{min}}$ MAE RMSE				
2D case:	4.52	9.63	29.01	203.3	42.73	67.83	Baseline	18543	72.62%	6992	27.38%	221191	99.35%	1453	0.65%
3D case:	0.30	0.56	9.32	66.85	3.83	13.05	FFNNs	24088	94.33%	1447	5.67%	222148	99.78%	496	0.22%
	1.56	2.13	12.88	95.96	13.67	30.85	KNN	18742	73.40%	6793	26.60%	217772	97.81%	4872	2.19%
	0.20	0.43	7.35	49.22	4.63	13.39	GBM	24914	97.57%	621	2.43%	214018	96.13%	8626	3.87%
	0.51	0.80	20.13	90.22	5.08	15.24	RF	23917	93.37%	1618	6.34%	212627	95.50%	10017	4.50%

*TP: True Positive FN: False Negative TN: True Negative FP: False Positive

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