

Cooperative Navigation for Large Swarms of Munitions in Three-Dimensional Flight

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Outline

- Range-only cooperative navigation
- Measurement schemes for formation flight
- MultiBoom High Fidelity Simulation
- Internal Models and Kalman Filters
 - Child Munition Guidance
 - Modified Linear Implementation
 - Centralized Navigation Kalman Filter
- Results
- Conclusions
- Extensions and work in progress























Intra-formation Measurement Schemes

• Undirected graph (directed later for sensitivities)





- Use a formation of 10 projectiles
- Represent edges of the graph as matrix sparsity pattern

1										
2	X									
3	X	X								
4	X	X	X							
5	X	X	X	x						
6		X	X	x	X					
7			X	x	X	x				
8				x	X	x	X			
9					X	X	X	X		
0						x	X	X	X	
	1	2	3	4	5	6	7	8	9	0

Scheme A



1										
2	X									
3	X	X								
4	X	X	X							
5		X	X	X						
6			X	X	X					
7				X	X	X				
8	X				X	X	X			
9	X	X				X	X	X		
0	X	X	X				X	X	X	
	1	2	3	4	5	6	7	8	9	0

Scheme B



	1										
	2										
	3	X									
	4	X	X								
ne (5		X	X							
าอน	6	X		X	X						
Scl	7	X	X		X	X		j.			
	8	X	X	X		X	X				
	9	X	X	X	X		X	X			
	0	X	X	X	X	X		X	X		
		1	2	3	4	5	6	7	8	9	0















MultiBoom High Fidelity Simulation

- Fortran 95 code using structures
- Bodies modeled in 6DOF with quaternions
- Bodies easily added to the formation
- Can add physical connections and aero surfaces
- Modules added:
 - Cooperative Navigation time update
 - Cooperative Navigation measurement update
 - Child EKF derivatives
 - LQR tracking control



MultiBoom modifications—"daisy chain"





Child munition functional block diagram





Internal Models: Child Guidance system



Internal Models: Child EKF

$$egin{aligned} \mathbf{K}_{ekf} &= \mathbf{P}_{ekf} \mathbf{H}_{ekf}^T \mathbf{R}_{ekf}^{-1} \ \mathbf{K}_{ekf} &= \mathbf{A}_{ekf} \mathbf{P}_{ekf} + \mathbf{P}_{ekf} \mathbf{A}_{ekf}^T + \mathbf{G}_{ekf} \mathbf{Q}_{ekf} \mathbf{G}_{ekf}^T - \mathbf{P}_{ekf} \mathbf{H}_{ekf}^T \mathbf{R}_{ekf}^{-1} \mathbf{H}_{ekf} \mathbf{P}_{ekf} \ \dot{\mathbf{x}}^+ &= \dot{\mathbf{x}}^- + \mathbf{K}_{ekf} (\mathbf{z} - \hat{\mathbf{z}}) \ \mathbf{H}_{ekf} &= \begin{bmatrix} \mathbf{I}_{5 imes 5} & \mathbf{0} \end{bmatrix} \ \mathbf{A}_{ekf} &= rac{\partial \mathbf{f}}{\partial \mathbf{x}} \end{aligned}$$



Parent Extended Kalman Filter





Internal Models: Parent EKF

$$\dot{\hat{x}_{i}}^{-} = V_{i}c_{\theta i}c_{\psi i}$$
$$\dot{\hat{y}_{i}}^{-} = V_{i}\psi_{i}c_{\theta i}$$
$$\dot{\hat{y}_{i}}^{-} = -V_{i}s_{\theta i}$$
$$\dot{\hat{z}_{i}}^{-} = -V_{i}s_{\theta i}$$
$$\dot{P}_{ekf} = \mathbf{G}_{ekf}\mathbf{Q}_{ekf}\mathbf{G}_{ekf}^{T}$$

 $\begin{aligned} \mathbf{K}_{ekf} &= \mathbf{P}_{ekf} \mathbf{H}_{ekf}^{T} \left(\mathbf{H}_{ekf} \mathbf{P}_{ekf} \mathbf{H}_{ekf}^{T} + \mathbf{R}_{ekf} \right)^{-1} \\ \mathbf{P}_{ekf}^{+} &= \left(\mathbf{G}_{ekf} - \mathbf{K}_{ekf} \mathbf{H}_{ekf} \right) \mathbf{P}_{ekf} \\ \hat{\mathbf{x}}^{+} &= \hat{\mathbf{x}}^{-} + \mathbf{K}_{ekf} (\mathbf{z} - \hat{\mathbf{z}}) \end{aligned}$



Where

$$\begin{aligned} \hat{\mathbf{z}}_{k} &= h_{k} = \hat{\rho}_{ij} \\ \rho_{ij} &= \sqrt{(x_{j} - x_{i})^{2} + (y_{j} - y_{i})^{2} + (z_{j} - z_{i})^{2}} \\ \mathbf{H}_{ekf}(k, 1 + 3(i - 1)) &= \frac{\partial h_{k}}{\partial \hat{x}_{i}} = -(\hat{x}_{j} - \hat{x}_{i})/\hat{\rho}_{ij} \\ \mathbf{H}_{ekf}(k, 2 + 3(i - 1)) &= \frac{\partial h_{k}}{\partial \hat{y}_{i}} = -(\hat{y}_{j} - \hat{y}_{i})/\hat{\rho}_{ij} \\ \mathbf{H}_{ekf}(k, 3 + 3(i - 1)) &= \frac{\partial h_{k}}{\partial \hat{z}_{i}} = -(\hat{z}_{j} - \hat{z}_{i})/\hat{\rho}_{ij} \\ \mathbf{H}_{ekf}(k, 1 + 3(j - 1)) &= -\mathbf{H}_{ekf}(k, 1 + 3(i - 1)) \\ \mathbf{H}_{ekf}(k, 2 + 3(j - 1)) &= -\mathbf{H}_{ekf}(k, 2 + 3(i - 1)) \\ \mathbf{H}_{ekf}(k, 3 + 3(j - 1)) &= -\mathbf{H}_{ekf}(k, 3 + 3(i - 1)) \end{aligned}$$



H_{ekf} sparsity, scheme A





H_{ekf} sparsity, scheme B





H_{ekf} sparsity, scheme C





Scenario





Results RMS estimation error SL – Scheme 5





Results – Scheme 2345





Results – Scheme 2468





Results—String Stability





Terminal Estimation Error Scheme 2468C

Noise Intensity	<i>e</i> _x [[m]	e_y	[m]	<i>e_z</i> [m]		
	min	max	min	max	min	max	
0%	0.3	23.3	0.02	18.34	0.13	12.1	
2%	0.6	26.1	0.86	21.9	0.47	14.8	
4%	0.4	23.6	0.11	18.57	0.08	12.0	
6%	0.5	24.4	0.21	19.4	0.06	12.3	
8%	0.6	24.3	0.26	19.3	0.03	12.2	



Conclusions

- Nine measurement topologies tested in a simulation
- Centralized EKF used with range-only measurements
- Wide distribution of measurements advantageous
 - Global vs. Local information
 - Similar to 'accordion' effect in long vehicle convoys



Future Work

- Improved sensor covariance model
- Investigate distributed filtering strategy
- Use slant range measurements directly in Child EKF
 - No intermediate estimation of 3D position coordinates
- A priori design of the measurement topology
 - Use Gauss PS guidance waypoints to choose
 - Find a metric to optimize topology
 - Optimize using GA
 - Using 30 of 45 possible intra-formation measurements

$$\binom{45}{30} = 344,867,425,584$$



Work in Progress

Parent-child swarming projectile concept





Trajectory Design for Parent / Child Munitions

- Solve a ballistic trajectory using a linearized subset of the MPLT eqns. where V and p are held constant, and $c_{\theta}=1$, $s_{\theta}=\theta$.
- Using the linear solution as an initial guess, solve a ballistic trajectory using the MPLT model with t_f held constant and $u_1=u_2=0$.
- Using the non-linear ballistic solution and $\lambda = \text{rand}, u_1 = \text{rand}, u_2 = \text{rand}, \text{ solve for the optimal controlled trajectory using all MPLT eqns.}$
- Repeat the last step for each child munition, using the parent munition solution as the initial guess. Note that only the target range and target crossrange need to be modified from previous



Toward a Measurement Topology Metric

• Test for non-defective Tetrahedron basis for each projectile





Test for non-defective tetrahedron

• The sextuple $(a, b, c, \tilde{a}, \tilde{b}, \tilde{c})$ is Facial

$$\min(a + b + c, a + \tilde{b} + \tilde{c}, \tilde{a} + b + \tilde{c}, \tilde{a} + \tilde{b} + c)$$

>
$$\max(a + \tilde{a}, \tilde{b} + b, \tilde{c} + c)$$

• The Calyer-Menger determinant is positive

$$\begin{vmatrix} 0 & 1 & 1 & 1 & 1 \\ 1 & 0 & a^2 & b^2 & c^2 \\ 1 & a^2 & 0 & \tilde{c}^2 & \tilde{b}^2 \\ 1 & b^2 & \tilde{c}^2 & 0 & \tilde{a}^2 \\ 1 & c^2 & \tilde{b}^2 & \tilde{a}^2 & 0 \end{vmatrix} > 0$$



log(.) Cayley-Menger determinant





log(.) Cayley-Menger determinant





Back up Slides





Results



Position with EKF estimates





Euler angles with EKF estimates





Shaping the Future of Aerospace