

# Conjunction Assessment Risk Analysis



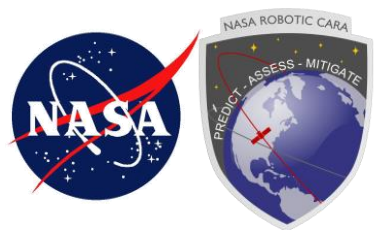
## Time Dependence of Collision Probabilities During Satellite Conjunctions

Doyle T. Hall<sup>1</sup>,  
Matthew D. Hejduk<sup>2</sup>, and  
Lauren C. Johnson<sup>1</sup>

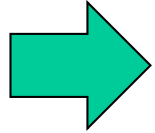
The 27<sup>th</sup> AAS/AIAA Space Flight Mechanics Meeting  
San Antonio TX, 2017 Feb 5-9

<sup>1</sup>Omitron Inc.

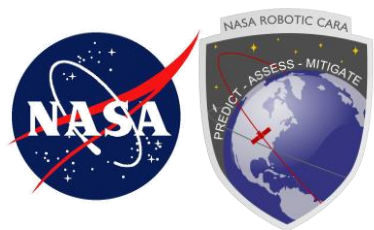
<sup>2</sup>Astrorum Consulting LLC



# Outline

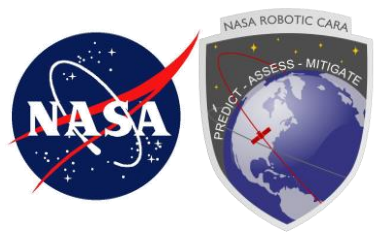


- **Motivation and objectives**
- **Overview of collision probability theory**
- **Analysis of well-studied conjunctions**
- **Analysis of archived conjunctions**
- **Conclusions**



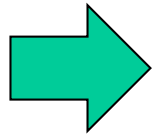
# Motivation and Objectives

- **Motivation:** The probability of collision,  $P_c$ , between two Earth-orbiting satellites can often but not always be approximated adequately using the “2D  $P_c$ ” formulation
- **Objective:** Implement an improved method to estimate collision probabilities
  - Use Coppola’s analytical “3D  $P_c$ ” formulation\*
  - Validate using well-studied test cases and Monte Carlo methods
  - Compare 2D and 3D  $P_c$  for archived conjunctions



# Outline

- Motivation and objectives



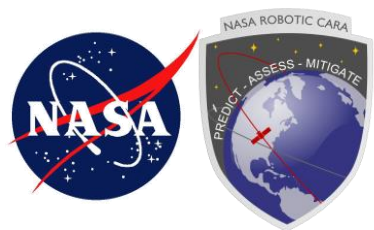
- **Overview of collision probability theory**

**Monte Carlo methods, 3D  $P_c$  theory, 2D approximations**

- Analysis of well-studied conjunctions

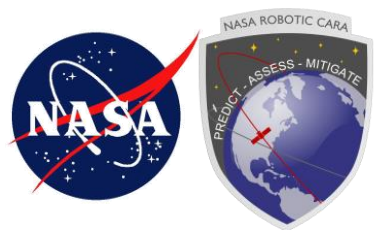
- Analysis of archived conjunctions

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# Monte Carlo $P_c$ Estimation

- **Collision probabilities can be estimated using Monte Carlo simulations**
  - **Computationally intensive, especially for low probability events**
- **Alfano\* analyzes twelve conjunctions in detail using Monte Carlo simulations**
  - **Benchmark test cases that can be used for validation of the 3D  $P_c$  software**
  - **Includes cases where the 2D  $P_c$  method both succeeds and fails**



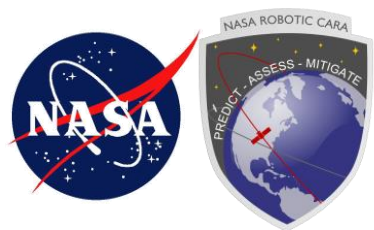
# Coppola's 3D $P_c$ Formulation

- Coppola\* provides an analytically-derived formulation to calculate  $P_c$  and its time derivative

$$P_c = P_0 + \int_{t_0}^{t_0+T} \left( \frac{dP_c}{dt} \right) dt \quad \text{1D time integral}$$

$$\frac{dP_c}{dt} = \oint_{4\pi} I(\hat{\mathbf{r}}, t) d^2\hat{\mathbf{r}} \quad \text{2D unit - sphere integral}$$

- These integrals must be calculated numerically



# Coppola's 3D $P_c$ Formulation

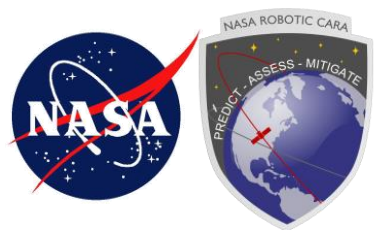
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**Analyzing the probability rate\* provides new insight into the time dependence of conjunction risks**

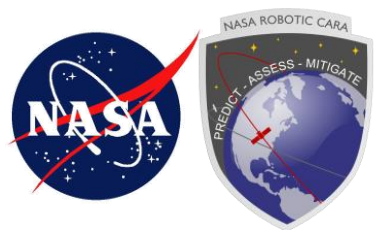
\*K.J.DeMars, Y.Cheng, and M.K.Jah (2014) "Collision Probability with Gaussian Mixture Orbit Uncertainty", *J. Guidance Control and Dynamics*, **37**(3) 979-985, 2014.



# Coppola's Conjunction Time Bounds

- **Coppola\*** also provides estimates for the bounding times of a conjunction
  - These often bracket the nominal time of closest approach (TCA), but not always
- These bounds only provide a first-cut approximation for the limits of the numerical integration over time
  - These limits sometimes need to be expanded to bracket sufficiently the time(s) when  $dP_c/dt$  peaks



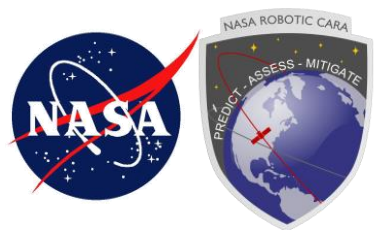


# CARA's Current 3D $P_c$ Implementation

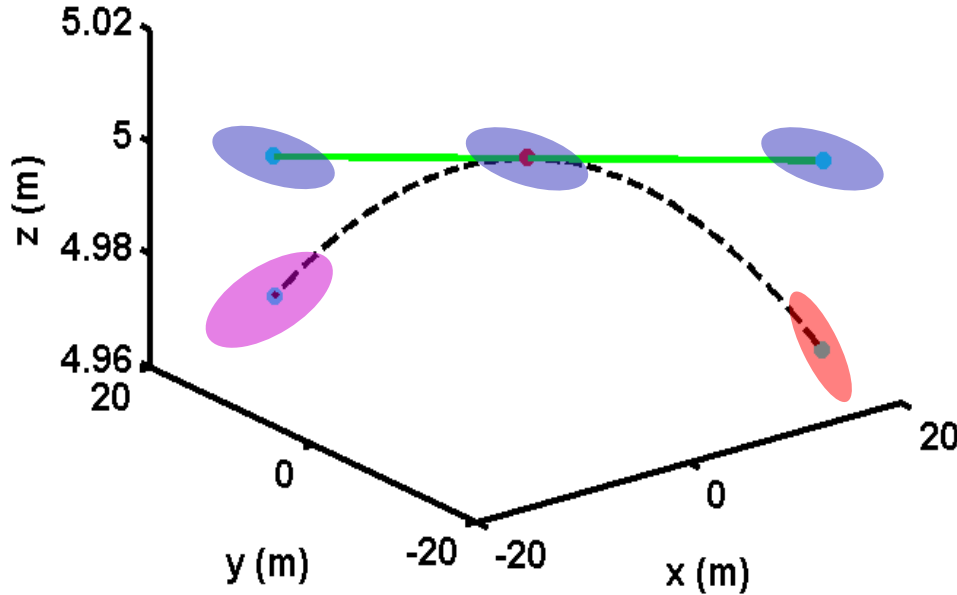
- The 3D  $P_c$  method is general enough to use
  - Gaussian Mixture Model state distributions\*
  - Complex dynamical motion models
  - Full 6x6 time-dependent state covariances
- CARA's current implementation uses
  - Single Gaussian ECI state distributions
  - The Keplerian two-body motion model
  - Full 6x6 ECI-state covariances, propagated using an analytically-derived state transition matrix#
- Future plans include more advanced approaches

\* J. T. Horwood, *et al.* (2011) "Gaussian Sum Filters for Space Surveillance: Theory and Simulations," *J. of Guidance, Control, and Dynamics*, vol.34, p.1839–1851.

# S.W. Shepperd (1985) "Universal Keplerian State Transition Matrix," *Celestial Mechanics*, vol.35, p.129-144.



# Schematic Illustration of the Encounter Region



*Illustration based on Alfano's\* test case #2*

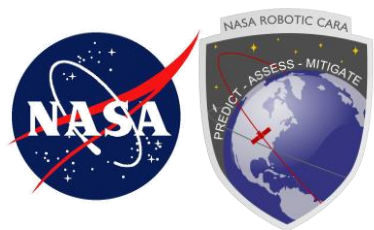
*NOTE: Actual  $1\sigma$  surfaces are much larger and thinner*

## 2D $P_C$ assumptions:

- Presumes straight trajectory (green)
- Presumes static covariances (blue)

## 3D $P_C$ assumptions:

- Trajectories are curvilinear (black)
- Covariances vary throughout the encounter (pink, orange)



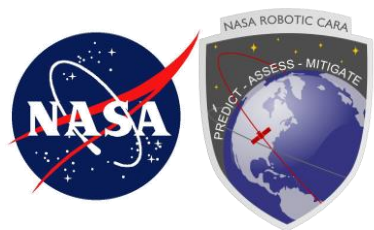
## Approximations Used for 2D $P_c$ Estimation

- In terms of the relative position/velocity state vector and the associated 6x6 covariance matrix:

$$\mathbf{x}(t) = \begin{bmatrix} \mathbf{r}(t) \\ \mathbf{v}(t) \end{bmatrix} \approx \begin{bmatrix} \mathbf{r}(t_{ca}) + (t - t_{ca})\mathbf{v}(t_{ca}) \\ \mathbf{v}(t_{ca}) \end{bmatrix}$$





$$\mathbf{P}(t) = \begin{bmatrix} \mathbf{A}(t) & \mathbf{B}(t)^T \\ \mathbf{B}(t) & \mathbf{C}(t) \end{bmatrix} \approx \begin{bmatrix} \mathbf{A}(t_{ca}) & \mathbf{0}_{3 \times 3} \\ \mathbf{0}_{3 \times 3} & \mathbf{0}_{3 \times 3} \end{bmatrix}$$

- Here  $t_{ca} = \text{TCA} = \text{the time of closest approach}$



# Step-by-Step Relaxation of the 2D $P_c$ Approximations

Using CARA's current 3D  $P_c$  software, the three approximations used in the 2D  $P_c$  method can be relaxed in a step-by-step manner:

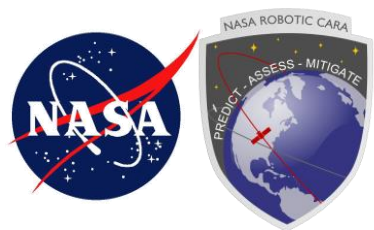
-  Coppola 1: Linear motion,  $A=A(TCA)$ ,  $B=C=0$
-  Coppola 2: Kep2Body,  $A=A(TCA)$ ,  $B=C=0$
-  Coppola 3: Kep2Body,  $A=A(t)$ ,  $B=C=0$
-  Coppola 4: Kep2Body,  $P=P(t)$

Step “Coppola 1” employs all of the 2D  $P_c$  assumptions

Step “Coppola 2” introduces Keplerian 2-body motion

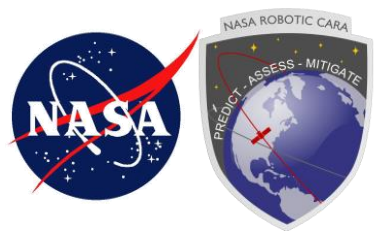
Step “Coppola 3” introduces time-varying position covariances

Step “Coppola 4” introduces position+velocity covariances



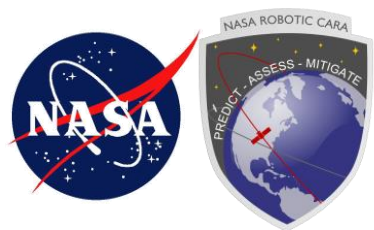
# Outline

- Motivation and objectives
- Overview of collision probability theory
- ➔ • **Analysis of well-studied conjunctions**  
    **Benchmark test cases analyzed by S. Alfano\***
- Analysis of archived conjunctions
- Conclusions



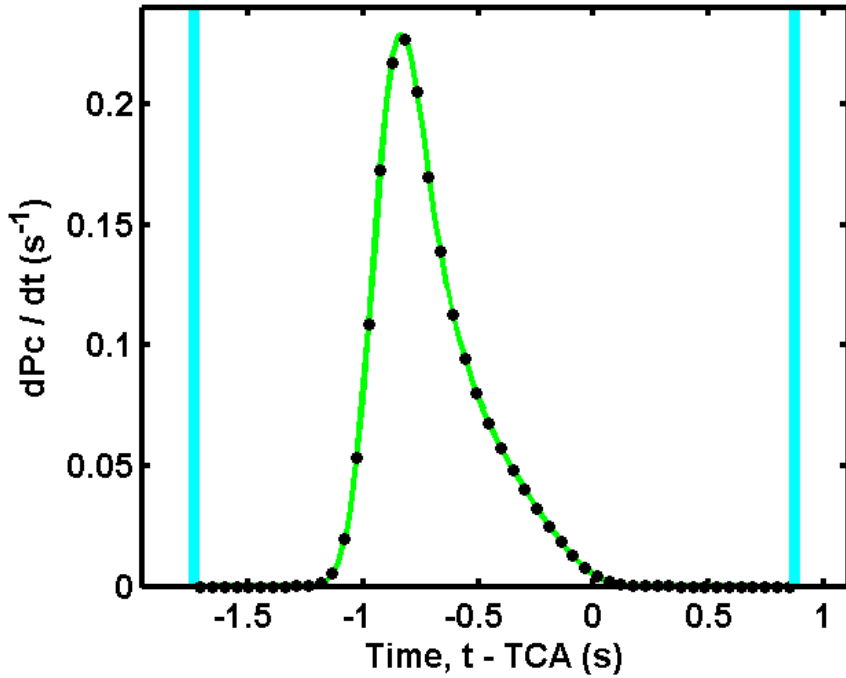
# Validation using Alfano's\* Benchmarks

- **Benchmark case #3:**
  - “Linear” case where 2D  $P_c$  is known to be accurate
  - The 3D  $P_c$  software correctly reproduces the 2D  $P_c$  approximation, and Alfano's benchmark  $P_c$  value
- **Benchmark case #10:**
  - “Nonlinear” case where 2D  $P_c$  is known to be inaccurate
  - The 3D  $P_c$  software correctly reproduces Alfano's benchmark  $P_c$  value
- **Other benchmark cases also analyzed (but not shown)**

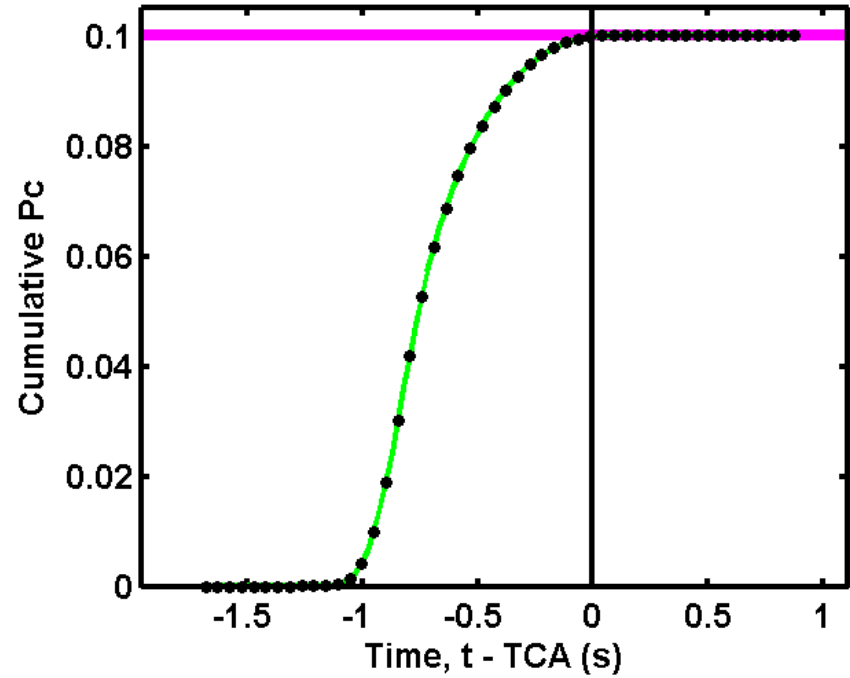


# Alfano's\* "Linear" Test Case #3

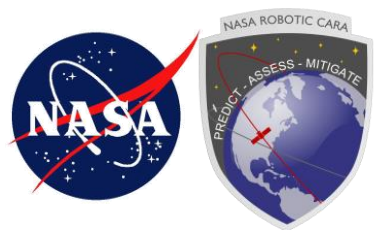
- Coppola bounds for  $\gamma = 1e-16$
- Coppola (linear motion)
- MC (linear motion,  $N = 3e+07$ )



- Foster  $P_c = 0.10035$
- Coppola  $P_c = 0.10034$
- MC  $P_c = 0.100352 \pm 0.000058$



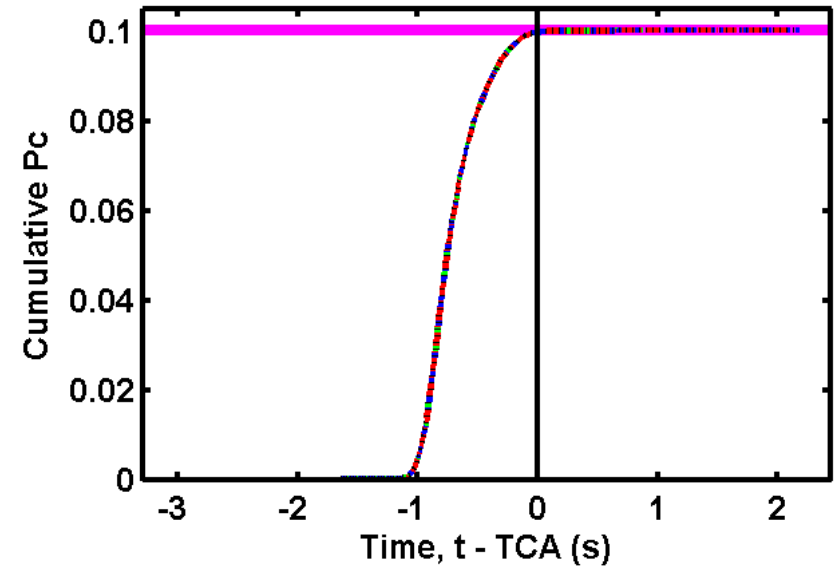
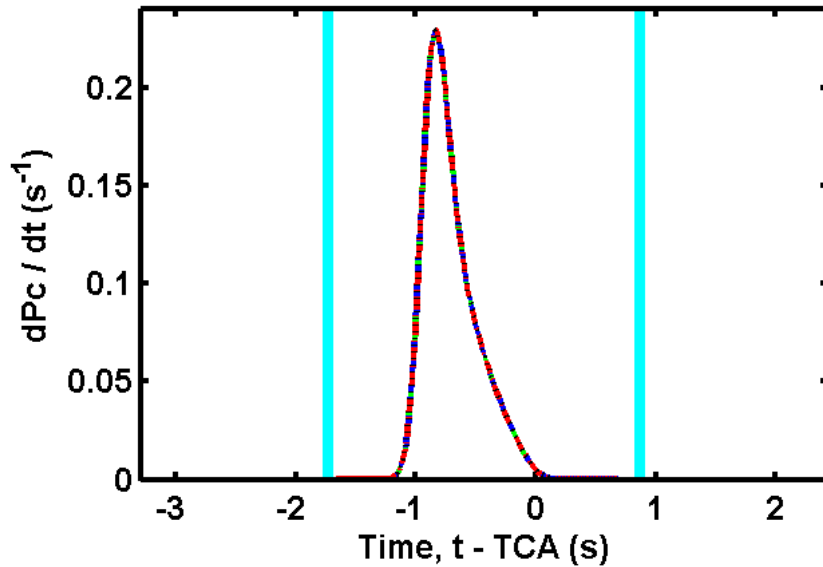
**These plots validate that the 3D  $P_c$  software correctly reproduces both the Monte Carlo and 2D  $P_c$  estimates, when using the 2D  $P_c$  approximations**



# Alfano's\* "Linear" Test Case #3

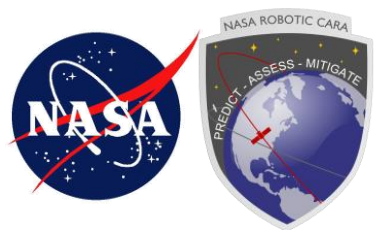
- Coppola bounds for  $\gamma = 1e-16$
- Coppola 1: Linear motion,  $A=A(TCA)$ ,  $B=C=0$
- - - Coppola 2: Kep2Body,  $A=A(TCA)$ ,  $B=C=0$
- . - . Coppola 3: Kep2Body,  $A=A(t)$ ,  $B=C=0$
- ..... Coppola 4: Kep2Body,  $P=P(t)$

- Foster 2D:  $P_c = 0.10035$
- Coppola 1:  $P_c = 0.10034$
- - - Coppola 2:  $P_c = 0.10034$
- . - . Coppola 3:  $P_c = 0.10034$
- ..... Coppola 4:  $P_c = 0.10034$



**These plots validate that the 3D  $P_c$  software correctly reproduces the 2D  $P_c$  estimate, even when the 2D  $P_c$  approximations are fully relaxed**

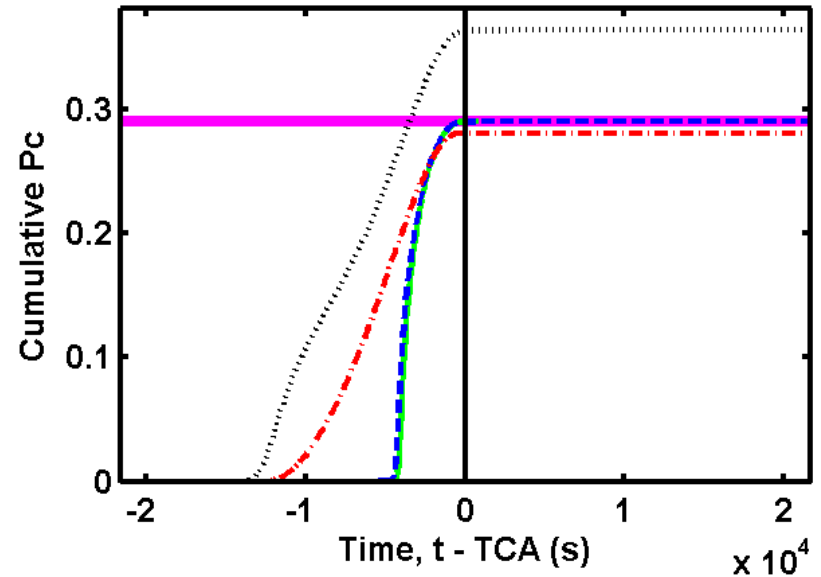
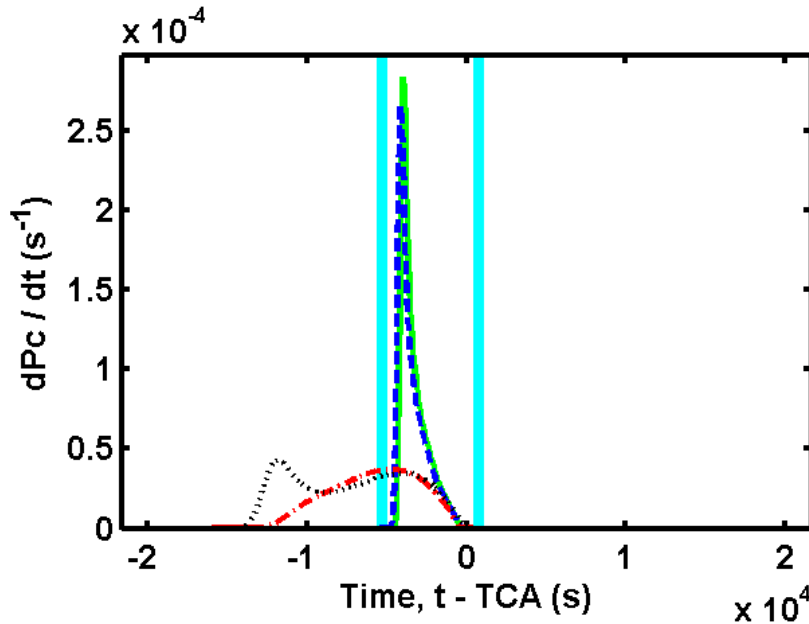




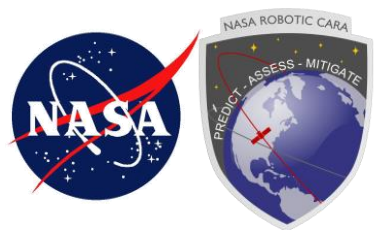
# Alfano's\* "Nonlinear" Test Case #10

- Coppola bounds for  $\gamma = 1e-16$
- Coppola 1: Linear motion,  $A=A(TCA)$ ,  $B=C=0$
- - - Coppola 2: Kep2Body,  $A=A(TCA)$ ,  $B=C=0$
- · - · - Coppola 3: Kep2Body,  $A=A(t)$ ,  $B=C=0$
- Coppola 4: Kep2Body,  $P=P(t)$

- Foster 2D:  $P_c = 0.29016$
- Coppola 1:  $P_c = 0.29016$
- - - Coppola 2:  $P_c = 0.28998$
- · - · - Coppola 3:  $P_c = 0.28108$
- Coppola 4:  $P_c = 0.36406$

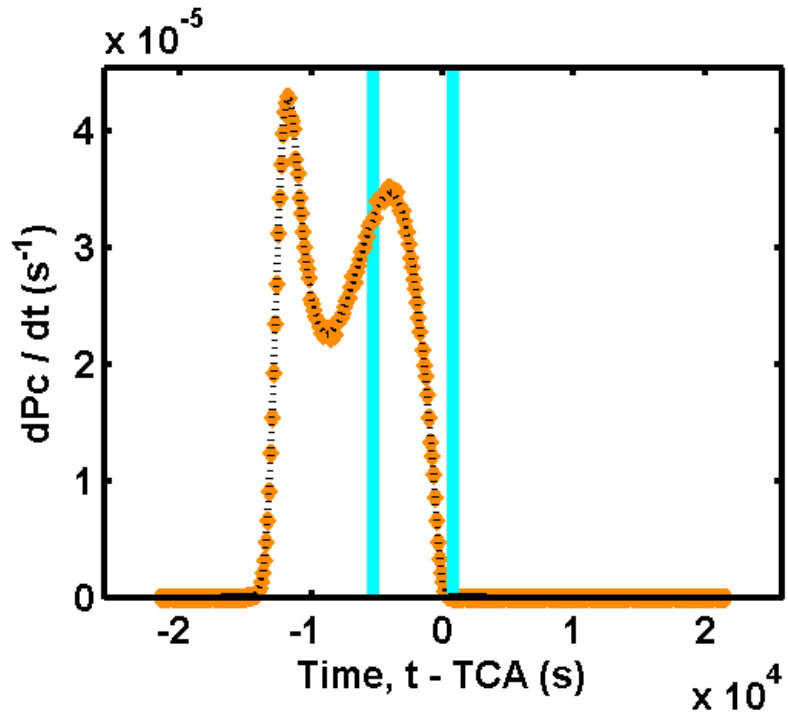


These plots validate that the 3D  $P_c$  software correctly yields different results as the 2D  $P_c$  approximations are relaxed in a step-by-step fashion

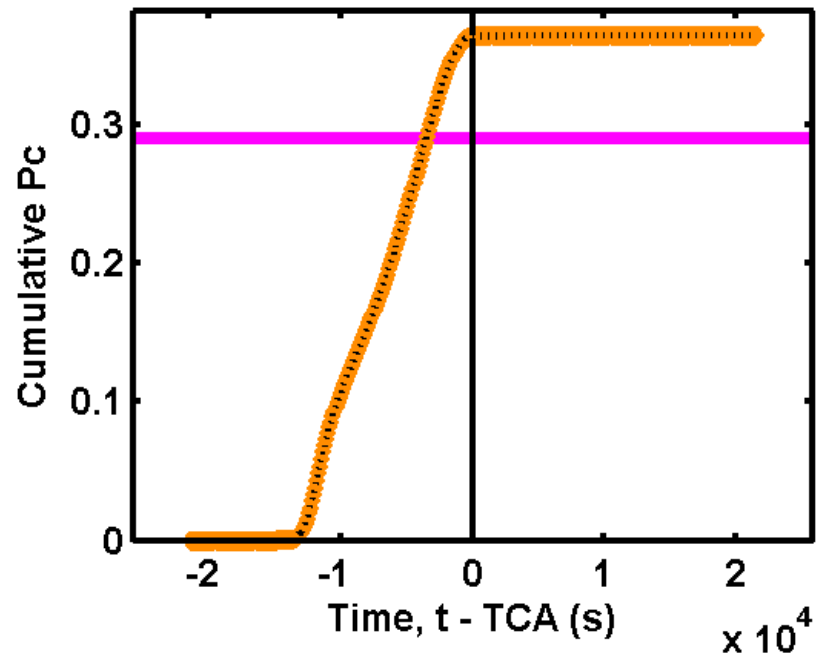


# Alfano's\* "Nonlinear" Test Case #10

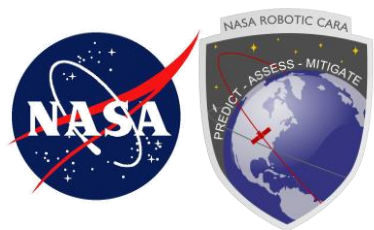
- Coppola bounds for  $\gamma = 1e-16$
- ◆ Monte Carlo (N = 3000000)
- ⋯ Coppola 4: Kep2Body, P=P(t)



- Foster  $P_c = 0.29016$
- ◆ MC  $P_c = 0.36380 \pm 0.00035$
- ⋯ Coppola  $P_c = 0.36406$

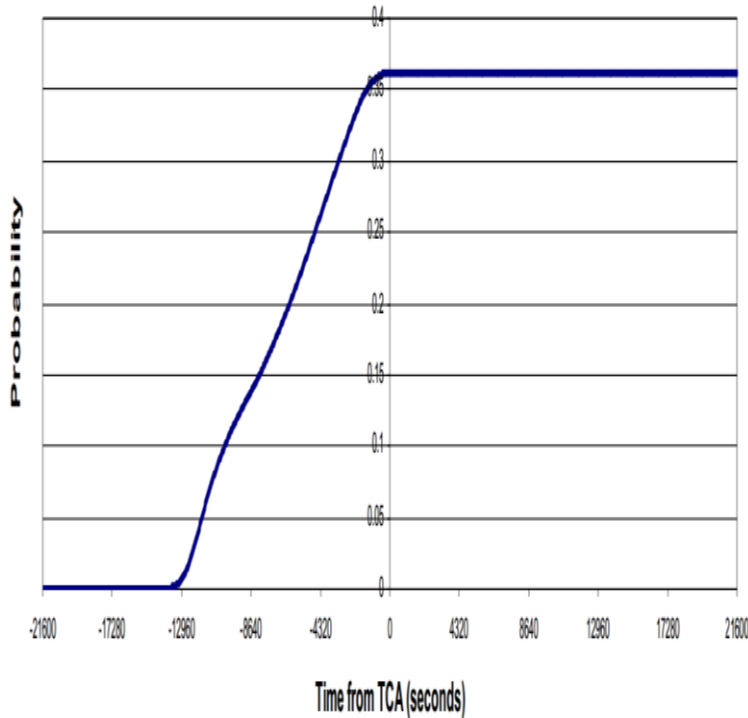


**These plots validate that the 3D  $P_c$  software correctly reproduces the Monte Carlo simulation, and that the  $dP_c/dt$  profile has two blended peaks**

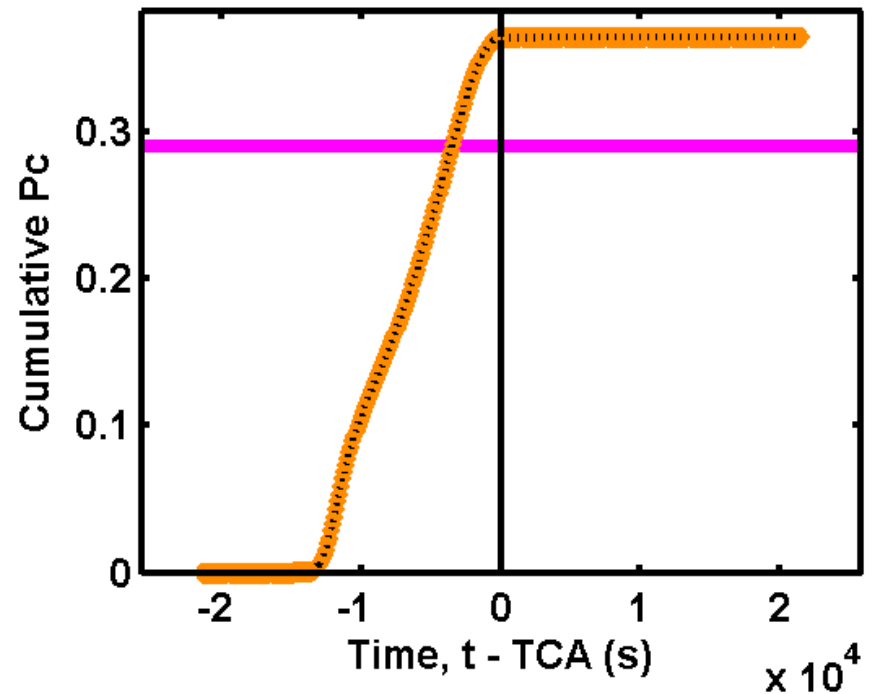


# Alfano's\* "Nonlinear" Test Case #10

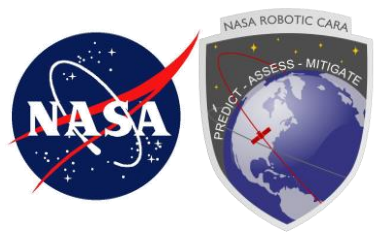
Alfano's\* original plot for cumulative probability



- Foster  $P_c = 0.29016$
- ◆ MC  $P_c = 0.36380 \pm 0.00035$
- ⋯ Coppola  $P_c = 0.36406$

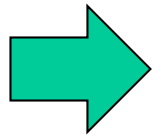


These plots validate that the 3D  $P_c$  software correctly reproduces Alfano's benchmark Monte Carlo results

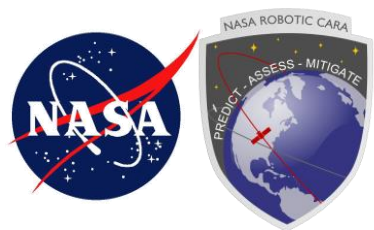


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- Motivation and objectives
- Overview of collision probability theory
- Analysis of well-studied conjunctions

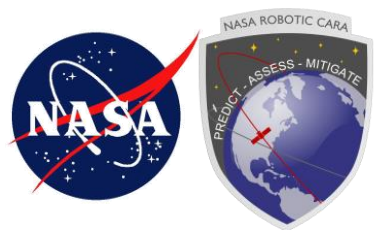


- **Analysis of archived conjunctions**  
2D vs. 3D results, repeating events, small- $P_c$  screening
- Conclusions



# Analysis of Archived Conjunctions

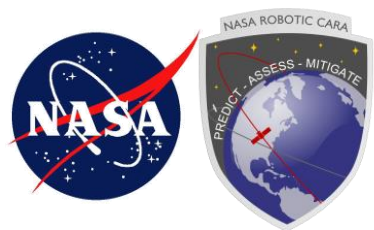
- **The 3D  $P_c$  method has been applied to 80,453 archived conjunctions**
  - Actual events that occurred between 2016 April 1 and 2016 June 1
- **Relatively few have appreciable 3D  $P_c$  values**
  - Only 11,211 (14%) have  $P_c \geq 10^{-15}$
  - Only 5,761 (7.2%) have  $P_c \geq 10^{-7}$
  - Only 2,674 (3.3%) have  $P_c \geq 10^{-5}$



# Analysis of Archived Conjunctions

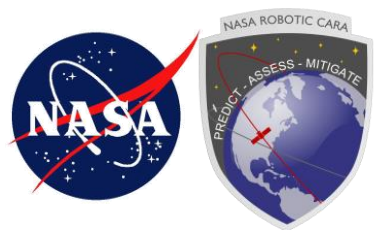
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**This is the most important set for the CARA team**



# Archived Conjunctions with 3D $P_c \geq 10^{-5}$

- For these, most 2D and 3D estimates were found to be relatively close to one another
  - 71% have 2D  $P_c$  and 3D  $P_c$  within 10% of one another
  - 85% have 2D  $P_c$  and 3D  $P_c$  within 30% of one another
- But smaller subsets were found to differ significantly
  - 5.6% have 2D  $P_c$  and 3D  $P_c$  separated by a factor of 3 or more
  - 2.4% have 2D  $P_c$  and 3D  $P_c$  separated by a factor of 10 or more
- The cases where 3D  $P_c \gg 2D P_c$  are of significant concern to the CARA team
  - Threatening conjunctions could be overlooked when using the 2D  $P_c$  approximation



# Repeating Events for Close Proximity Satellites

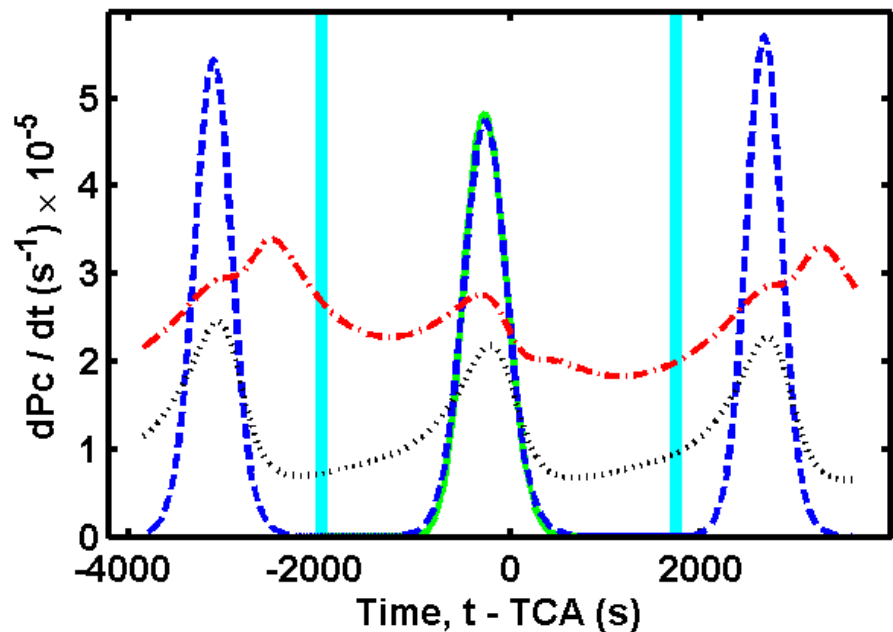
- Objects persistently orbiting in close proximity can make repeated close approaches to one another

- Satellites within formations or clusters
- These conjunctions can often be identified by their long durations
- This can create multiple, blended peaks in  $dP_c/dt$

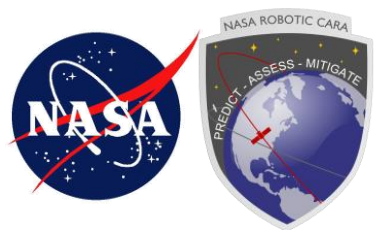
- These types of conjunctions explain some but not all of the archived cases that have  $3D P_c \gg 2D P_c$

## Archived conjunction involving two satellites flying in close proximity

- Coppola bounds for  $\gamma = 1e-16$
- Coppola 1: Linear motion,  $A=A(TCA)$ ,  $B=C=0$
- - - Coppola 2: Kep2Body,  $A=A(TCA)$ ,  $B=C=0$
- . - . Coppola 3: Kep2Body,  $A=A(t)$ ,  $B=C=0$
- ..... Coppola 4: Kep2Body,  $P=P(t)$



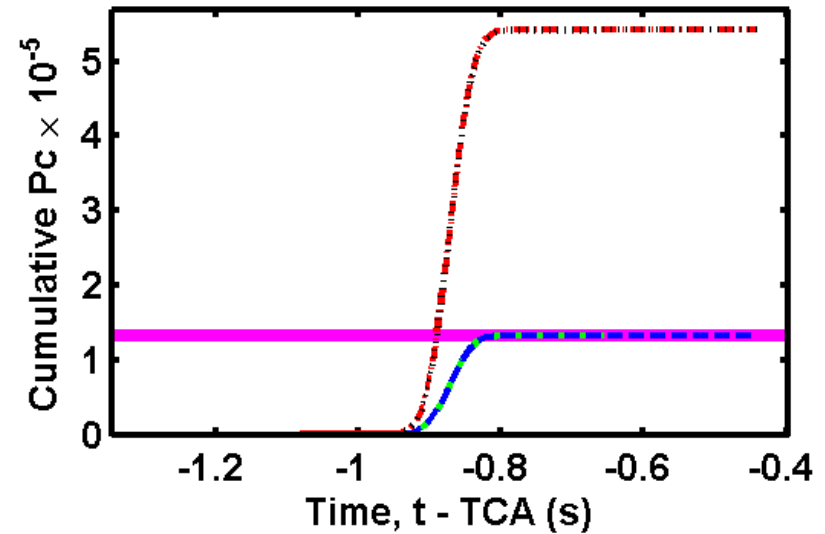
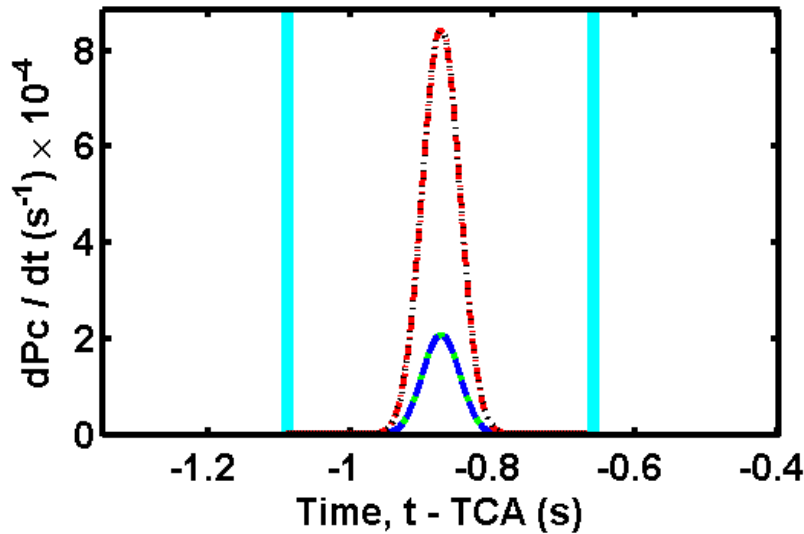




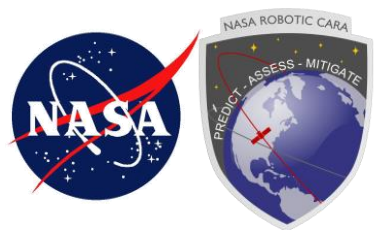
# Isolated Conjunctions with 3D $P_c \gg 2D P_c$

- Coppola bounds for  $\gamma = 1e-16$
- Coppola 1: Linear motion,  $A=A(TCA)$ ,  $B=C=0$
- - - Coppola 2: Kep2Body,  $A=A(TCA)$ ,  $B=C=0$
- . - . Coppola 3: Kep2Body,  $A=A(t)$ ,  $B=C=0$
- ⋯⋯⋯ Coppola 4: Kep2Body,  $P=P(t)$

- Foster 2D:  $P_c = 1.32804e-05$
- Coppola 1:  $P_c = 1.32808e-05$
- - - Coppola 2:  $P_c = 1.32809e-05$
- . - . Coppola 3:  $P_c = 5.42875e-05$
- ⋯⋯⋯ Coppola 4:  $P_c = 5.41930e-05$

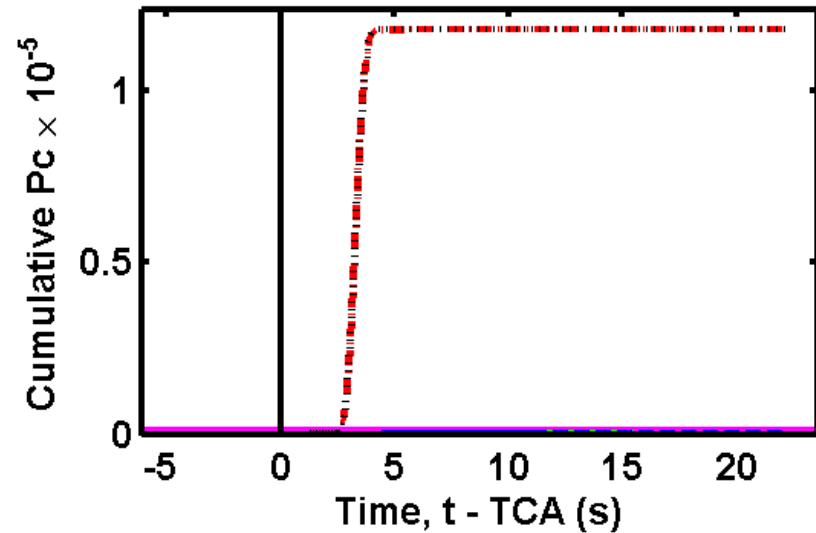
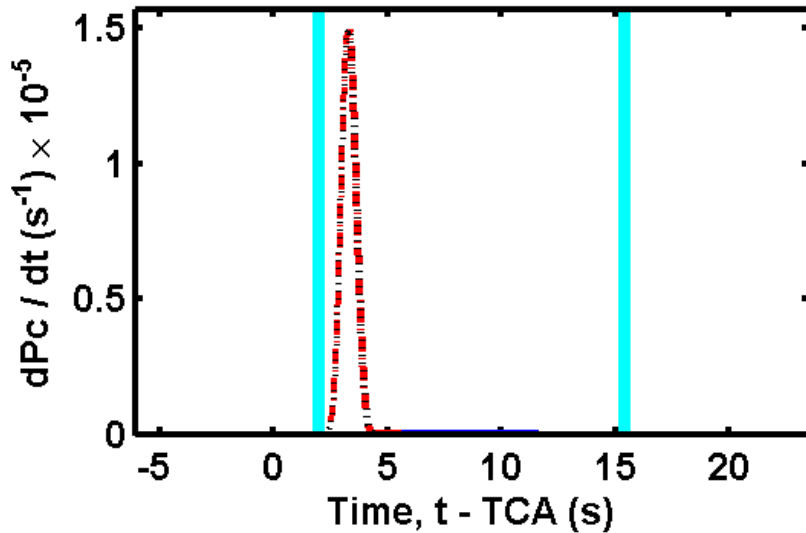


**Archived conjunction where the 3D  $P_c$  estimate exceeds the 2D  $P_c$  estimate by a factor of about four.**

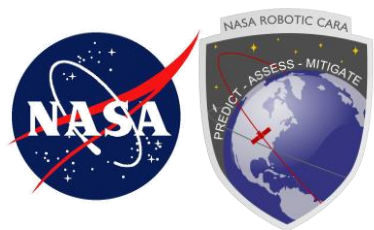


# Isolated Conjunctions with 3D $P_c \gg 2D P_c$

- Coppola bounds for  $\gamma = 1e-16$
  - Coppola 1: Linear motion,  $A=A(TCA)$ ,  $B=C=0$
  - Coppola 2: Kep2Body,  $A=A(TCA)$ ,  $B=C=0$
  - Coppola 3: Kep2Body,  $A=A(t)$ ,  $B=C=0$
  - Coppola 4: Kep2Body,  $P=P(t)$
- Foster 2D:  $P_c = 3.64637e-15$
  - Coppola 1:  $P_c = 3.64641e-15$
  - Coppola 2:  $P_c = 3.60376e-15$
  - Coppola 3:  $P_c = 1.17803e-05$
  - Coppola 4:  $P_c = 1.17956e-05$



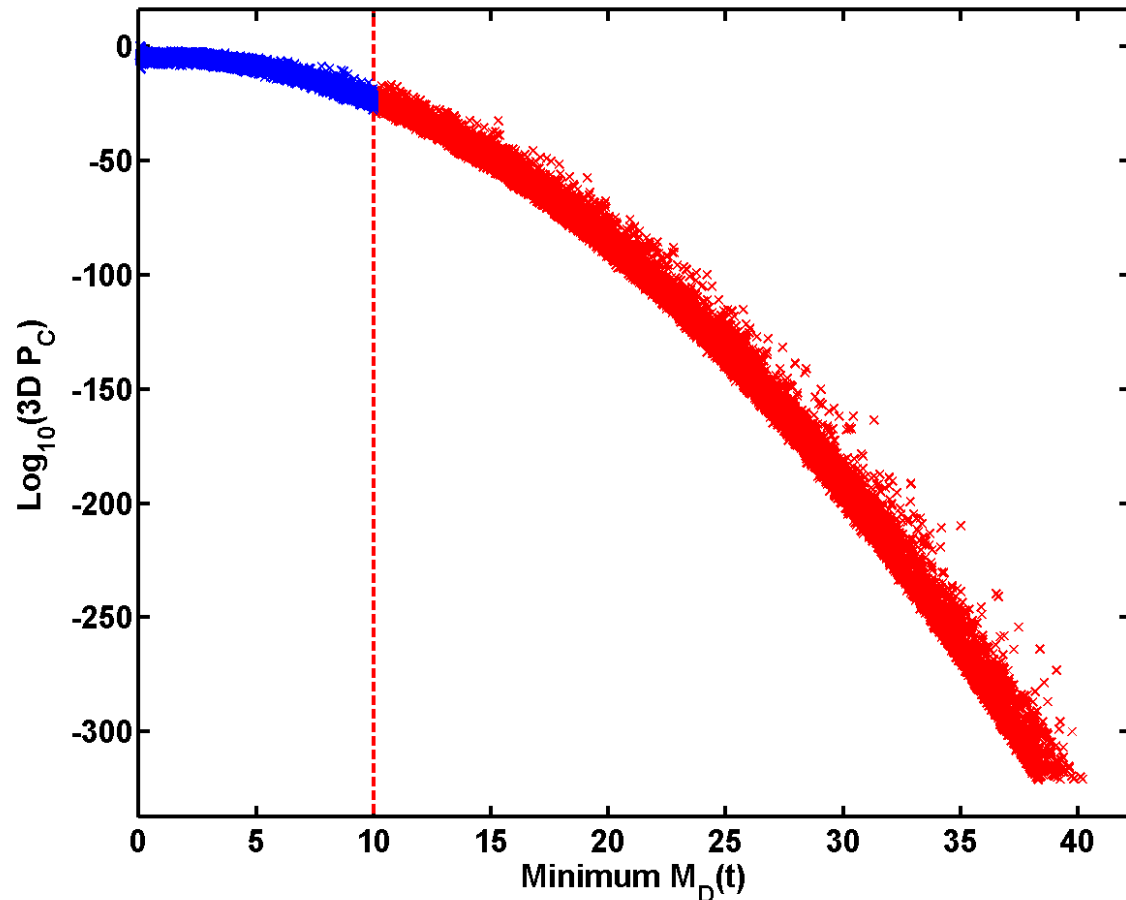
Archived conjunction where the 3D  $P_c$  estimate exceeds the 2D  $P_c$  estimate by several orders of magnitude.

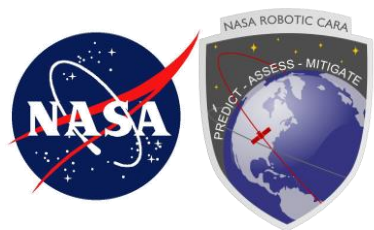


# A Screening Test for Small- $P_c$ Values

- Conjunctions with large relative-position Mahalanobis distances have small 3D  $P_c$  values
- This correlation provides the basis for an efficient small- $P_c$  screening test
- Applying this screening test eliminates the need to calculate 3D  $P_c$  for  $\approx 80\%$  of all conjunctions

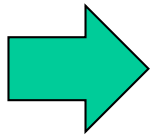
About 80% of the archived conjunctions have  $(M_D)_{\min} > 10$  and  $3D P_c < 3 \times 10^{-17}$

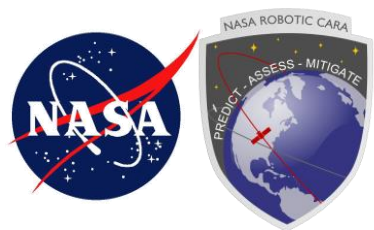




# Outline

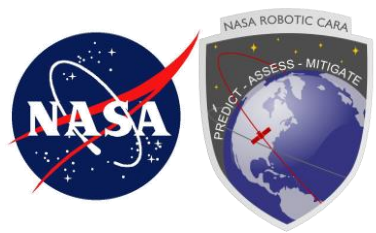
- **Motivation and objectives**
- **Overview of collision probability theory**
- **Analysis of well-studied conjunctions**
- **Analysis of archived conjunctions**
- **Conclusions**



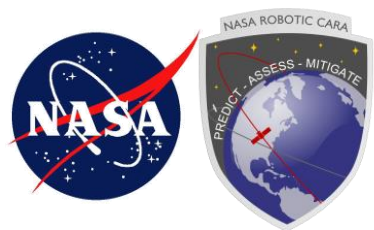


# Conclusions

- **The CARA team has implemented Coppola's 3D  $P_c$  formulation into software**
  - Validated using Alfano's benchmark test cases
  - Provides estimates for both  $P_c$  and  $dP_c/dt$
  - Provides insight into the time dependence of risk
- **Archived conjunction analysis indicates that**
  - Occasionally the 2D  $P_c$  approximation can be very inaccurate
  - An efficient small- $P_c$  screening test can be used to speed processing for large numbers of conjunctions

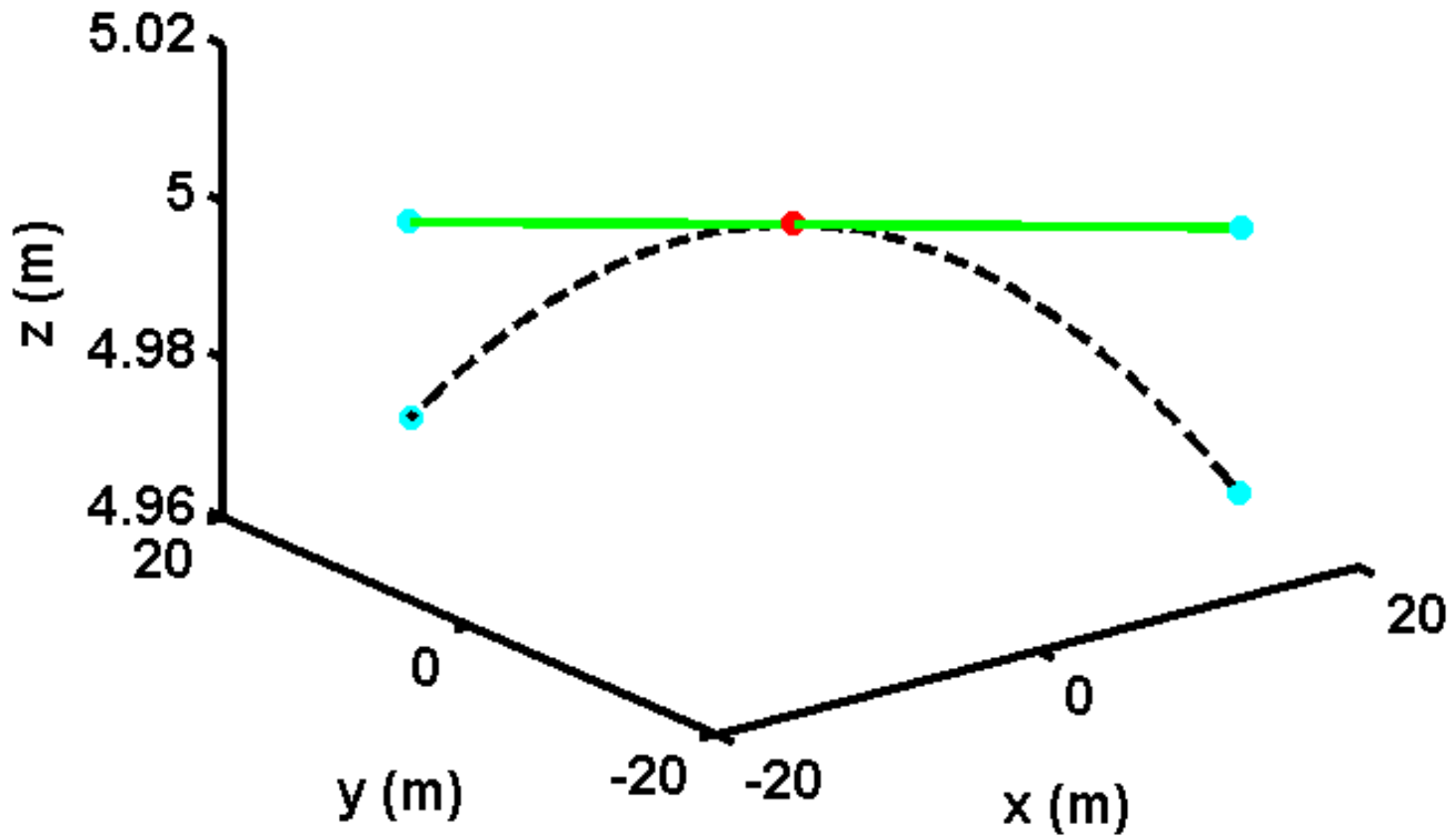


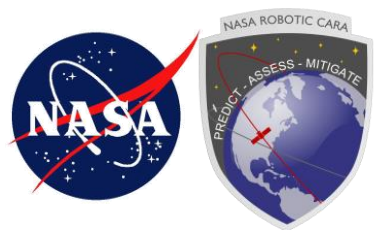
# Backup Slides



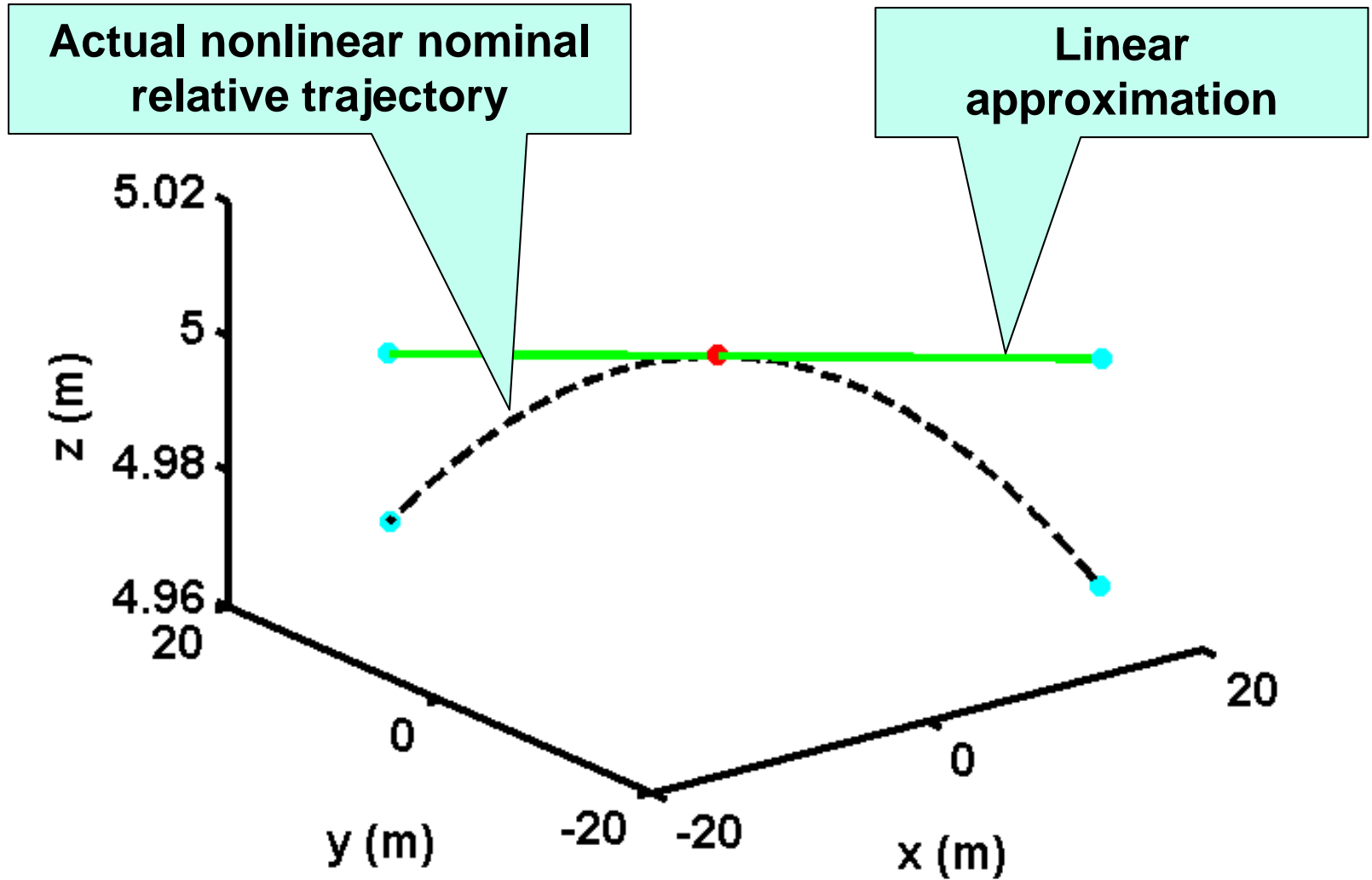
# Schematic Illustration of 2D $P_c$ Assumptions

Illustration of relative position trajectories for Alfano's (2009) "nonlinear" example #2

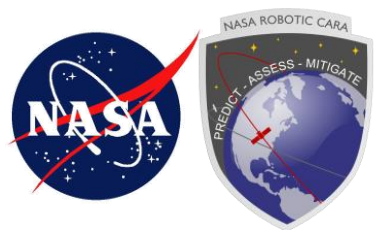




# Schematic Illustration of 2D $P_c$ Assumptions

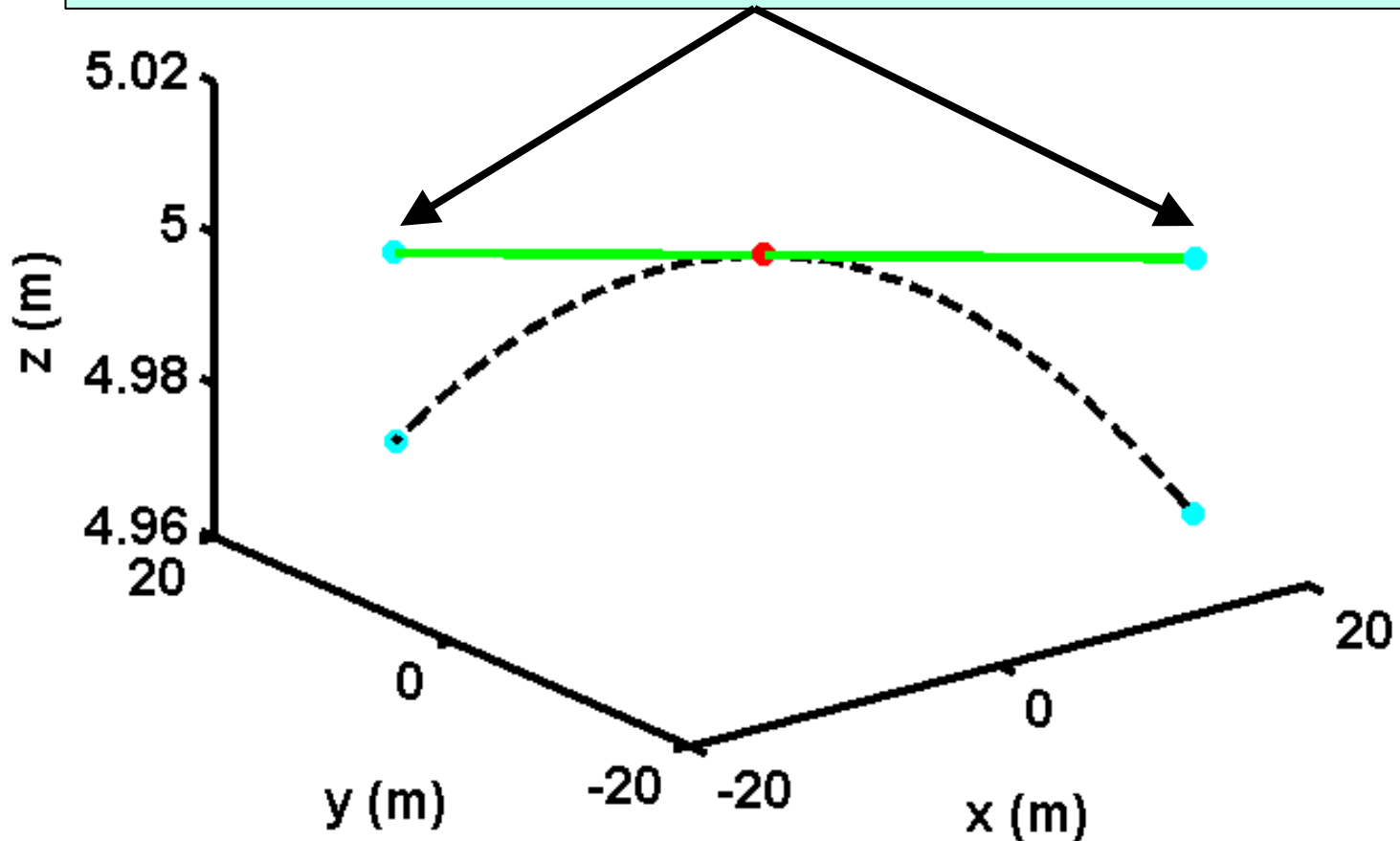


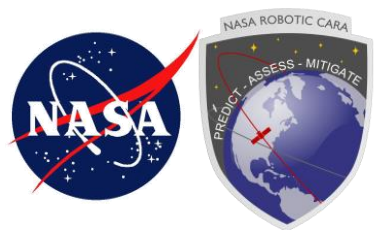




# Schematic Illustration of 2D $P_c$ Assumptions

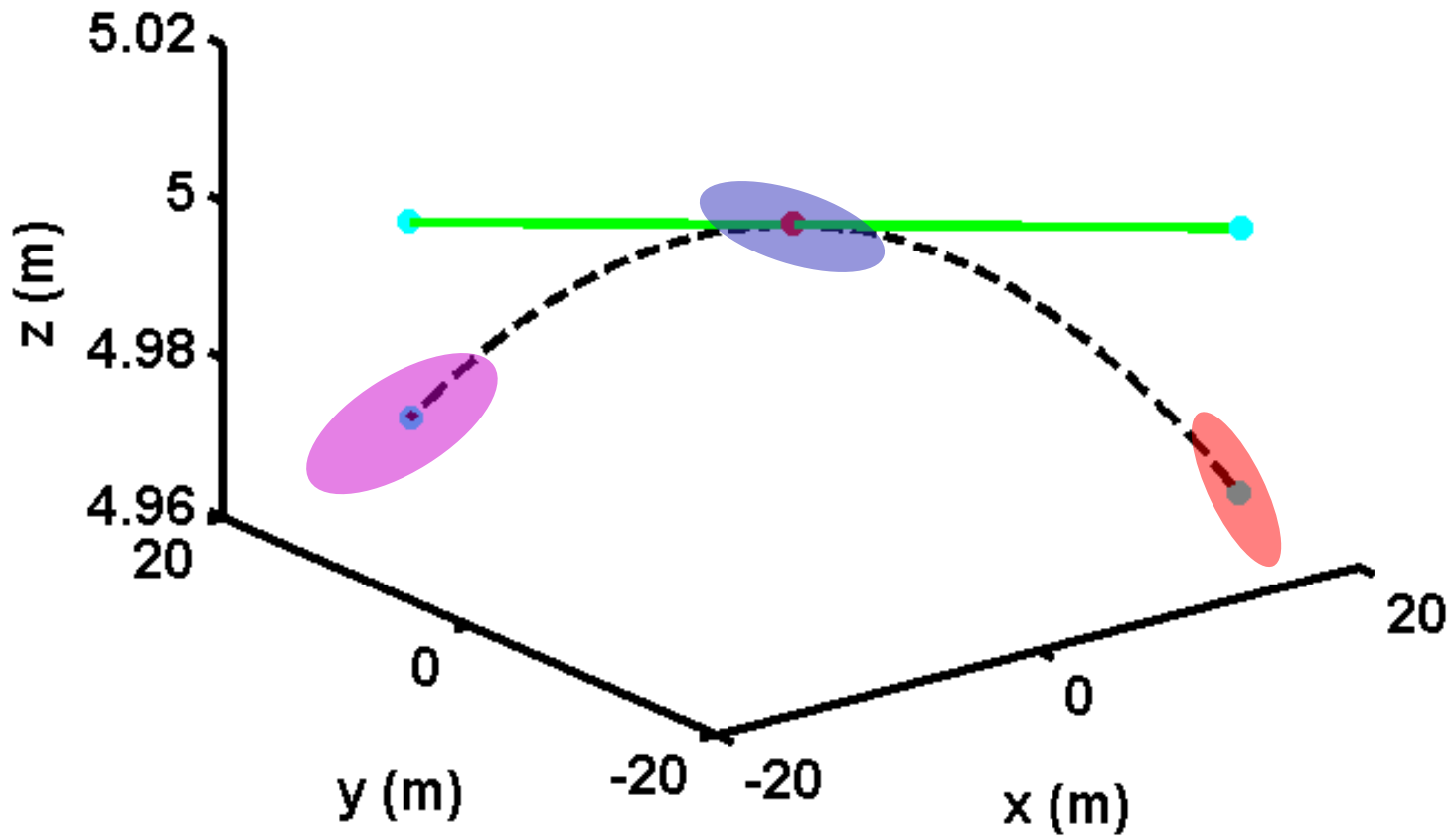
Light blue dots show the nominal relative positions at Coppola's conjunction time bounds

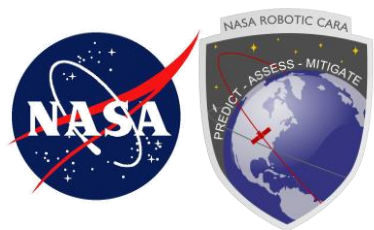




# Schematic Illustration of 2D $P_c$ Assumptions

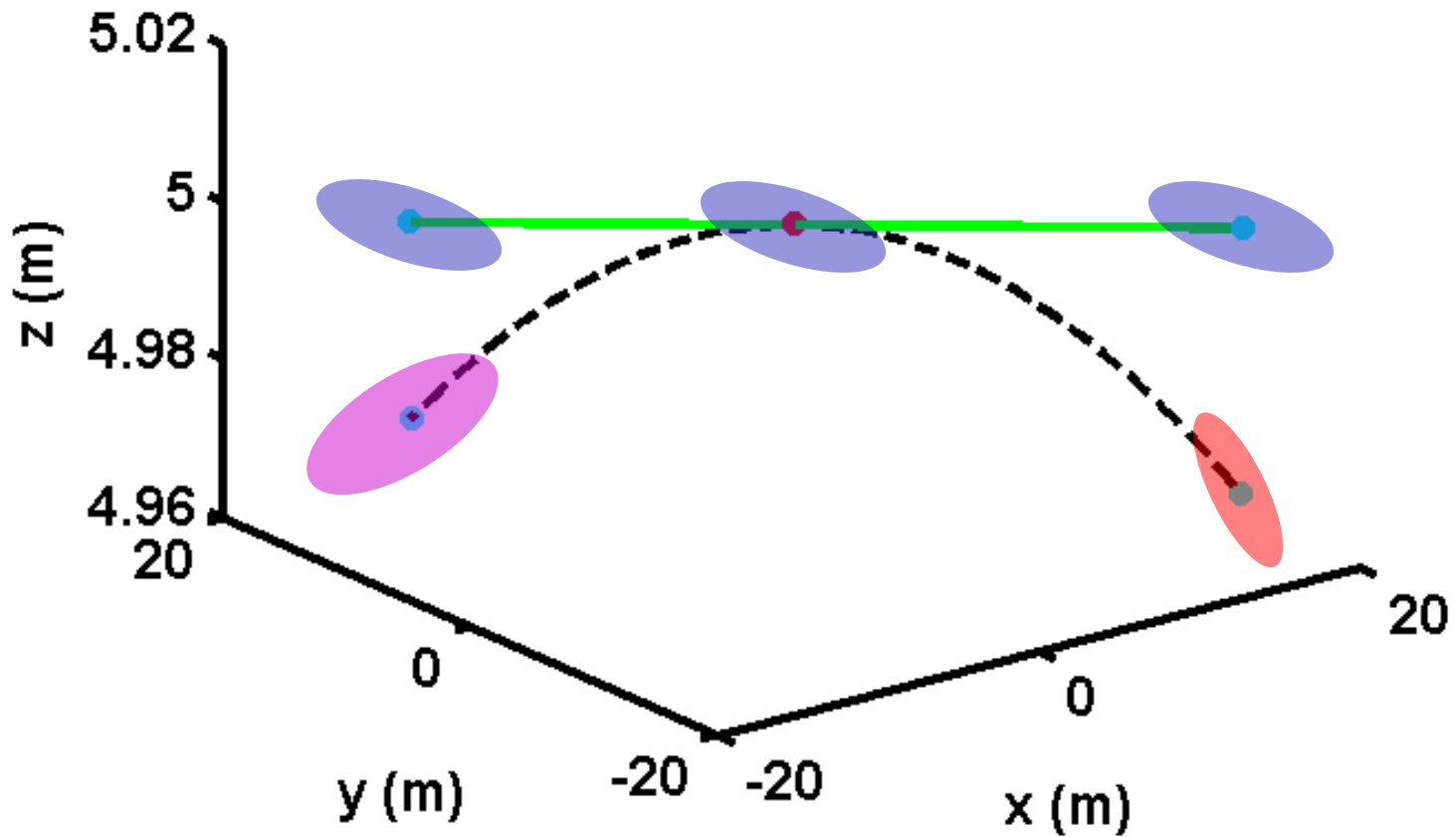
Relative position PDFs evolve in time  
**NOTE: Actual  $1\sigma$  surfaces are *much* larger and thinner**

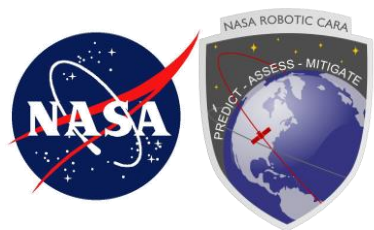




# Schematic Illustration of 2D $P_c$ Assumptions

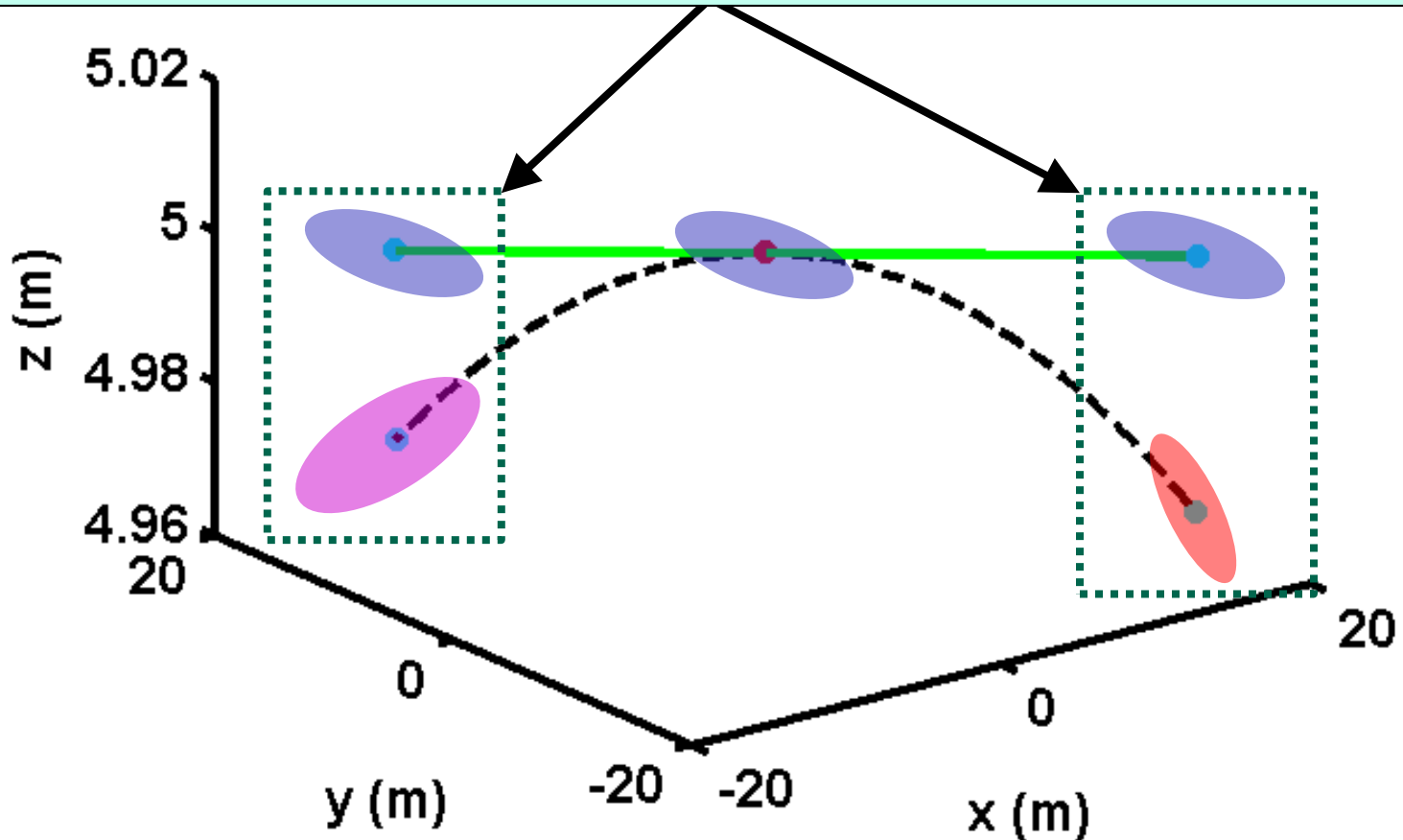
2D  $P_c$  approximates the PDFs as constant, and places them along the linearized trajectory

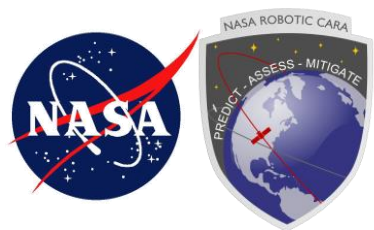




# Schematic Illustration of 2D $P_c$ Assumptions

The 2D  $P_c$  approximation will be inaccurate if these PDF differences become too large during the conjunction





# Mahalanobis Distance

The *Mahalanobis Distance* measures the difference between the positions of the primary and secondary objects, relative to the scale of their combined covariance:

$$M_D(t) = \left( \mathbf{r}^T \mathbf{A}^{-1} \mathbf{r} \right)^{1/2}$$

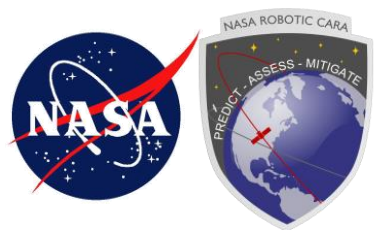
where

$$\mathbf{r} = \mathbf{r}(t) = \mathbf{r}_s - \mathbf{r}_p$$

(relative position)

$$\mathbf{A} = \mathbf{A}(t) = \mathbf{A}_s + \mathbf{A}_p$$

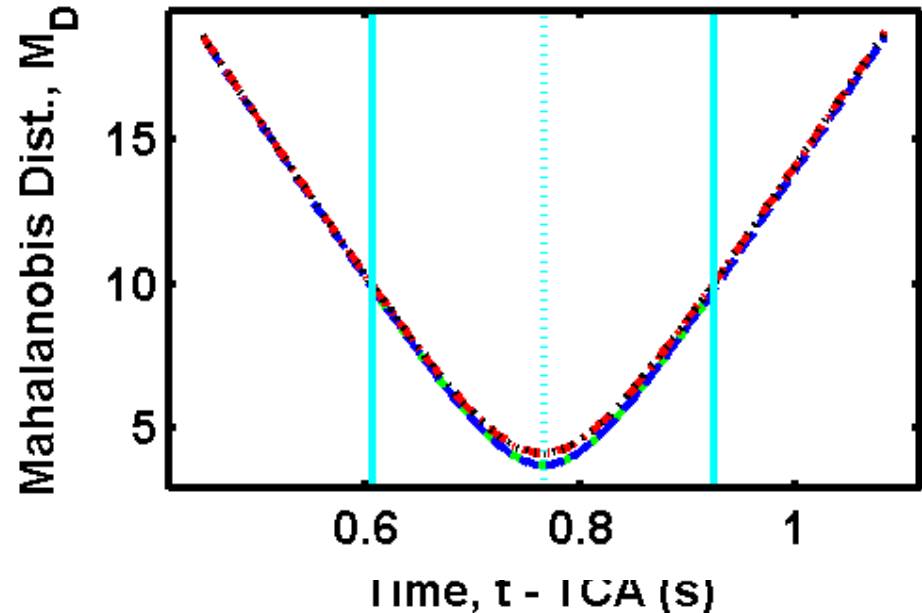
(combined covariance)

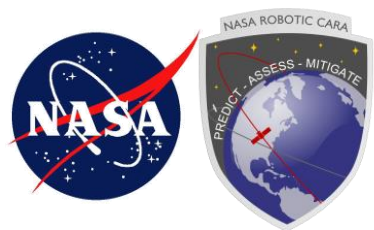


# Minimum Mahalanobis Distance

- The Mahalanobis distance varies as a function of time during a conjunction
- The minimum value  $(M_D)_{\min}$  often occurs near the conjunction midpoint, but not always
- $(M_D)_{\min}$  values vary significantly for different conjunction events

- Coppola bounds for  $\gamma = 1e-16$
- Coppola 1: Linear motion,  $A=A(TCA)$ ,  $B=C=0$
- - - Coppola 2: Kep2Body,  $A=A(TCA)$ ,  $B=C=0$
- . - . Coppola 3: Kep2Body,  $A=A(t)$ ,  $B=C=0$
- ..... Coppola 4: Kep2Body,  $P=P(t)$





# Monte Carlo $P_c$ Estimation Procedure

- 1. Sample the state PDFs for both the primary and secondary satellites**
- 2. Propagate the sampled states over the desired time span, checking if the separation becomes less than the combined hard-body radii**
- 3. If so, register a collision at the time the spheres defined by the hard-body radii make first first contact**
- 4. Repeat steps 1-3 to improve statistical estimation accuracy**