Filter Tuning Using the Chi-Squared Statistic

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Introduction

- The importance of a realistic covariance has grown in recent years due to its inclusion in conjunction analysis.
- The GSFC FDF has been responsible for providing the definitive orbit solutions for the Aqua and Aura missions for over a decade.
 - The orbit solutions utilizes Doppler data from ground stations and coherent range and Doppler from the NASA Space Network.
- The NASA Aqua and Aura missions have requested the covariance data be delivered, so that it can be used by the NASA conjunction assessment (CA) team.
- This prompted an analysis of the covariance coming out of the operational extended Kalman filter (EFK).
- Definitive and predictive accuracy of the orbit solutions remains an important priority for supporting science operations.



The Chi-squared Statistic

- The Chi-squared value is calculated in an attempt to characterize how well the covariance conforms to the actual error of a state.
- There are two inputs involved in the calculation
 - Error in the state
 - Calculated by comparing the predicted state to the definitive state
 - $\varepsilon = [\varepsilon_R \ \varepsilon_I \ \varepsilon_C]$
 - Covariance values
 - Position covariance taken directly from Orbit Determination Toolkit (ODTK) EKF values and converted to the RIC frame
 - C
- The chi-squared value is calculated as follows

$$- \chi_{3 dof}^2 = \varepsilon C^{-1} \varepsilon^T$$

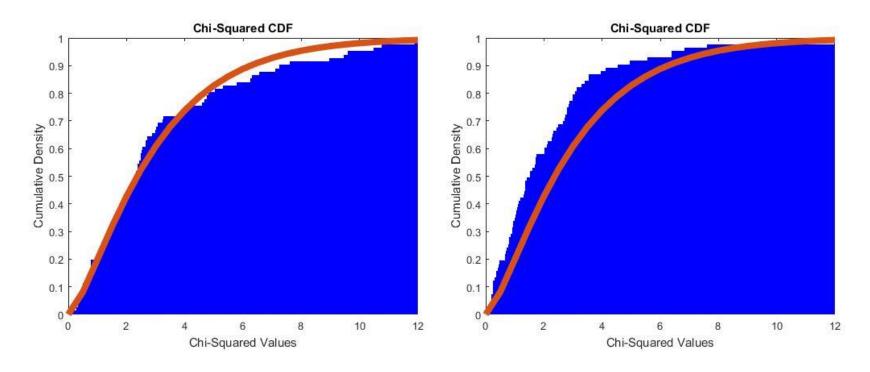


The Chi-squared Statistic (cont'd)

- In an ideal case, the error in the predicted state at any given point will match the error represented in the covariance matrix.
- For the three degree of freedom case of a satellite's orbit, the chisquared should have an average value of 3 and a standard deviation of 2.333.
- Average values greater than 3 indicate the covariance is not capturing the actual error in the propagation, while values less than 3 indicate that the covariance tends to be larger than the actual error
- Detailed treatment of the practical application of this statistic is available in the technical memorandum "Covariance Realism Evaluation Approaches" published by M.D. Hejduk



Baseline Scenario



Aqua Chi-squared Average: 3.29 Aqua Chi-squared Standard Deviation: 3.26

Aura Chi-squared Average: 2.23 Aura Chi-squared Standard Deviation: 2.41



Tuning Process

 The tuning process focused on adjusting the EKF parameters in order to improve both the realism of the covariance, as well as the predictive accuracy of the solution.

Methods used include

- Adjusting the white noise sigma values associated with tracking data measurements
- Implementing updated drag models
- Injecting process noise into the scenario to inflate the covariance

The tuning took place in two phases

- The first phase focused on improving the predictive accuracy of the solution
- The second focused on injecting process noise to appropriately size the covariance while retaining improvements in predictive accuracy



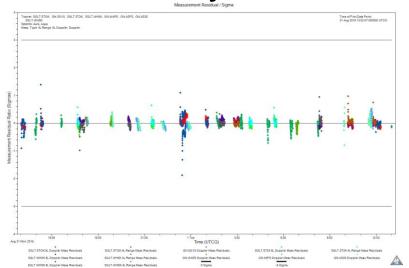
Tuning Predictive Accuracy

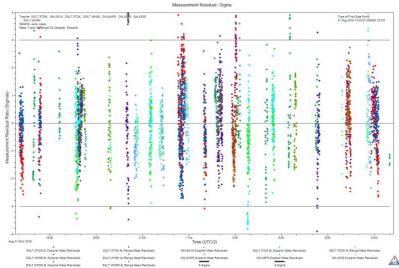
- The white noise sigma values associated with tracking data measurements were drastically reduced
 - The white noise values associated with the ground station Doppler measurements were dropped from 5 cm/sec to an average value of 0.5 cm/sec.
 - Each station was tuned individually
 - Shrinking the error associated with the SN measurements proved more complicated due to the fact that ODTK uses the actual covariance of the TDRS spacecraft as an input to its noise calculations
 - ODTK allows the noise to be reduced based on the final collection point of the SN measurement, but the majority of the noise in the system existed in the state uncertainty of the TDRS
 - Reducing the TDRS uncertainty involved tuning the ground based measurements used to determine the TDRS orbit state.



Tuning Predictive Accuracy (cont'd)

 The final result was a significant reduction in excess measurement noise in the filter





- Implementing updated drag models
 - A box and wing area plugin was implemented for both Aqua and Aura spacecraft assuming constant sun pointing for the solar panels
 - The result was actually a degradation in predictive accuracy. Further investigation is needed into why the modeling did not perform as expected



Injecting Process Noise

- Measurement tuning resulted in chi-squared average values over 8 for both spacecraft.
- To counteract this, measurement noise values were increased but doing this also appreciably degraded predictive accuracy.
- The focus turned to injecting process noise during propagation to effectively inflate the covariance.
- Further simplifying the equation for the chi-squared statistic can supply insight into the behavior in the individual components
 - Assume that the off-diagonal term of the covariance are zero and the equation can be reduced to

$$\chi_{3 dof}^2 = \frac{\varepsilon_R^2}{\sigma_R^2} + \frac{\varepsilon_I^2}{\sigma_I^2} + \frac{\varepsilon_C^2}{\sigma_C^2}$$



Injecting Process Noise (cont'd)

- From this equation it can be deduced that to achieve a chi-squared value of three, the predictive error and sigma values should have a one to one relationship.
- Process noise then should be injected with the goal being that the average sigma value matches the average predictive error in a certain component.
- To this effect, tuning runs were evaluated on a component basis in the radial, in-track and cross-track components
 - If the distribution of the error divided by the sigma ("Error over Sigma", or EOS) was 60 percent or higher, favoring either the predictive error or the sigma value, the process noise in that component was adjusted.
 - Analysis runs were completed until all of the components met this threshold for both spacecraft.

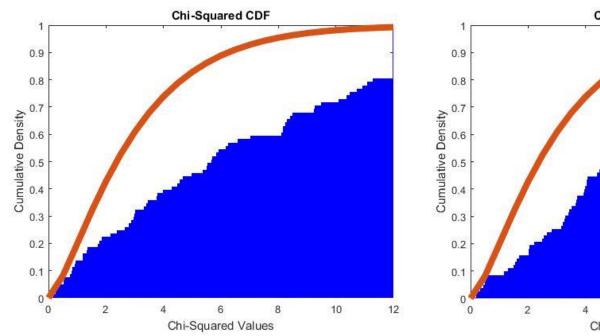


Results

Run	Satellite	Process Noise Injected (cm/sec)	Prediction Error (meters)	Chi-Squared Average
1	Aqua	AT: 0.075, CT: 0.001	185	8.77
	Aura	AT: 0.075, CT: 0.001	202	6.50
2	Aqua	AT: 0.060, CT: 0.001	185	8.77
	Aura	AT: 0.075, CT: 0.002	202	6.50
3	Aqua	AT: 0.060, CT: 0.003	191	8.2
	Aura	AT: 0.075, CT: 0.005	203	5.56
4	Aqua	AT: 0.075, CT: 0.300	179	6.59
	Aura	AT: 0.075, CT: 0.400	204	4.84
5	Aqua	AT: 0.050, CT: 0.300	181	6.63
	Aura	AT: 0.060, CT: 0.400	189	5.19
6	Aqua	AT: 0.030, CT: 0.300	193	7.59
	Aura	AT: 0.030, CT: 0.400	185	5.79



Results (cont'd)



Chi-Squared CDF

1
0.9
0.8
0.7
0.6
0.3
0.2
0.1
0.2
Chi-Squared Values

Aqua Chi-squared Average: 7.59

Aura Chi-squared Average: 5.79



Conclusions

- The process noise injection process applied did not produce an acceptable level of covariance realism.
 - Upon inspection it was clear that while the average predictive error and sigma values were close to a one to one relationship, the standard deviation of the predictive error was much higher than the standard deviation of the sigma value.
- Multiple metrics should be used to determine the effectiveness of a filter
 - The baseline scenario shows a good covariance realism but it is aided by measurement noise that is inflated to higher than realistic levels.
 - The runs with tightened measurement noise show better predictive accuracy but the chi-squared values show the covariance is not capturing the error in the propagated state.



Conclusions

- The dynamics of the problem play an important role in determining the key filter metrics
 - Both Aqua and Aura predictive error is heavily dominated by the in-track direction. This made tuning of the other components difficult due to the small nature of the values in those components.
 - It also places extra importance of effectively tuning the covariance in the intrack component, since it is the dominant source of error.



Future Work

- Investigate the impact of further tuning the process noise to capture the high standard deviation in the predictive error.
 - Tighten the thresholds on the EOS value so the process noise can be more effectively tuned
- Investigate techniques that would allow the covariance to scale appropriately with the predictive error standard deviation.
 - Possibly inject some process noise based upon the input to the atmospheric model
- Determine how the state of the solar cycle effects the performance of the filter in both predictive accuracy and covariance realism



Thank you

Questions?

