

The University of North Carolina at Chapel Hill Department of Computer Science

SCAAT Tracking: Numerical Error and Stability Concerns

Greg Welch



The Problem



Estimation in General

***** The scientific aspect

- Control and estimation theory
- Man-made and natural systems
- Systematic methods
- 🏶 The numerical aspect
 - Stochastic approach
 - Uses a computer numerically
 - KF developed for a computer
 - Affects of design and numerical error

III The Kalman Filter

Predictor-Corrector

- Seminal paper by R.E. Kalman, 1960
- Set of mathematical equations
- Optimal estimator (min. mean-square error)
- Estimation, filtering, prediction, fusion
- Predictor-corrector
- Recursive









Use the Dynamic Model



error covariance



Predict the Measurement (with measurement model and predicted state)





Compute Measurement Residual

image plane





Correct State and Covariance state = predicted + Kalman gain x residual





R = measurement uncertainty

Incorporates a notion of

direction of measurement information



III The Kalman Filter

www.cs.unc.edu/~welch

- Dissertation (appendix)
- Kalman filter web page



SCAAT Video

The SCAAT Kalman Filter

***** Constraints == Equations

- Single constraint == single equation
- Compare with Gauss-Seidel
 - Standard vs. SCAAT
 - Jacobi vs. Gauss-Seidel
 - Successive vs. simultaneous displacment
 - But stochastic not deterministic!



System Dynamics

- Use dynamic model
 Predict state
- ***** Predict error covariance



Geometric Interpretation

Project state unertainty into measurement space

uses Jacobian

• error magnification viewpoint

Combine with measurement uncertainty

added uncertainty
 Re-project normalized term
 Ratio (weight)



With Actual Measurement

Predict measurement

- Iower dimension than state
- incomplete information
- 🌾 Measure
- 🏶 Compute residual
- ***** Correct state
- * Correct error covariance



SCAAT Error and Stability

(† denotes special concern)



Some Sources & Effects

* Measurement error

- improper model structure
- non-white, non-normal
- derivation depends on it!

🏶 Dynamic error

- improper model structure
- improper parameters
- * Linearization error
 - an extension to the KF



Roundoff (Steady State)

Problem

- P must remain positive definite
- No driving noise...semidefinite...

Solutions

- High-precision operations
- Avoid completely deterministic systems
- Factor P (square-root, U-D)



Modeling Error

Problem

- You model a random constant
- Actually a random ramp

🍀 Solutions

- System identification is hard!
- Analyze residuals
- A priori system knowledge
- Minimize estimate interval



Observability †

Observability test
 (Controllability test)
 Local vs. global observability



General Conditions

- Uniformly completely observable †
 Bounded dynamic & measurement noise models
- 🏶 Bounded dynamic behavior



Complete Conditions

See equations 5.2 and 5.3 Bounded dynamics over time • Q finite but !=0 **Bounded measurement noise** • Reasonable for systems of interest **Sufficient constraints over time** • met by design or at run-time • Ironically SCAAT helps (fast) Sufficient sample rate †



A ctual Error







Steady-State Stability

Steady-state not always reachedKF is a linear operator

- transforms inputs into outputs
- transfer function
- Characteristic function
 - denominator of transfer function
 - roots provide information about stability
 - discrete: within unit circle in z plane
 - continuous: left half of s plane

General Application SCAAT and Systems of Equations

* Similar to Gauss-Seidel

- successive vs. simultaneous
- always use latest estimate
- Trade-off accuracy for work
 - Single constraints until "certain"
 - P matrix indicates certainty
- Small noise for stability



The University of North Carolina at Chapel Hill Department of Computer Science

SCAAT Tracking: Numerical Error and Stability Concerns

Greg Welch