### A Sufficient Comparison of Trackers

David Bizup Systems and Information Engineering University of Virginia dfb2ph@virginia.edu Donald Bizup Systems and Information Engineering University of Virginia brown@virginia.edu

## Need for Sufficient Comparisons

- Peculiar deficiency in the literature, trackers judged on
  - one or two measures of performance
  - one or two trajectories
- Fair method, based on sufficient comparisons:
  - Define comprehensive set of maneuvers
  - Define relevant measures of performance
  - Generate simulated radar data for every trajectory
  - Run data through each tracker
  - Compare error distributions via sufficiency

### **Trajectories and Measures**

- 15,120 constant rate turns:
- Measures of Performance
  - Position accuracy
  - Speed accuracy
  - Heading accuracy
  - Range accuracy
  - Bearing accuracy
  - Range rate accuracy
  - Cross-range rate accuracy

Trajectory Parameter	Levels
Range at start of maneuver	20, 40,, 300 miles
Initial heading	0, 30,, 330  degrees
Speed	200, 300,, 800 knots
Centripetal Acceleration	0, 1,, 5 g
Turn Direction	left, right

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### Trackers

- Two mode interacting multiple model
  - 1 nonmaneuvering mode
  - 1 maneuvering mode
  - Standard for tracking maneuvering tarets
  - Bar-Shalom and Li, *Multitarget-Multitarget Tracking: Principles and Techniques*, YBS (1995)
- Single mode, adaptive c<sub>min</sub> Kalman filter
  - Use range rate to produce statistic of accelerations (7<sup>th</sup> ICCRTS)
  - Gain set based on the statistic
  - Simpler than IMM
  - Uses additional information

## **Statistical Sufficiency**

- Probabilistic sufficiency is analagous to more familiar statistical sufficiency
  - If A is sufficient for  $\theta$ , then B contains no more information about  $\theta$  than A
  - *B* can be interpreted as *A*, plus noise
- Analogy is imperfect
  - *A* is either a sufficient statistic, or not
  - Concept of minimally sufficient statistic
  - Sufficiency of *A* does not depend on other statistics

# **Probabilistic Sufficiency**

Blackwell, *Equivalent Comparisons of Experiments*, Annals of Mathematical Statistics, vol. 24, pp265-272 (1953)

- Relationship between estimators
  - Either ASB, or BSA, or neither, or both
  - Possible that *CSA* and *CSB*
  - Establishes a partial order
- Definition: ASB if B is a stochastic transformation of A
  - *B* can be interpreted as *A*, plus noise
- Powerful, because if ASB, then A is better than B
  - A has a lower Bayes risk for <u>any</u> prior, <u>any</u> loss function

### Informativeness

[**Definition: Informativeness**] It is said that estimator A is more *informative* than estimator B, denoted AIB, if  $R_A \leq R_B$  for all prior distributions and all loss functions, where  $R_i$  is the Bayes Risk of estimator *i*.

# If *A* is more informative than *B*, then *A* is better than *B* no matter how better is defined!

Unfortunately, no constructive way to determine informativeness.

## Sufficiency

**[Definition: Sufficiency]** It is said that estimator A is *sufficient* for estimator B, denoted ASB, if there exists a stochastic transformation  $\psi : B \times A \to \text{Re}^+$  such that the following are satisfied where  $\omega$  denotes the predict or true state of nature.

$$\begin{array}{lll} f_B\left(b|\omega\right) &=& \int_A \psi\left(b|a\right) f_A\left(a|\omega\right) da, \, \omega \in \Omega, \, b \in \mathcal{B} \\ \psi\left(b|a\right) &\geq& 0 \\ \int_B \psi\left(b|a\right) db &=& 1 \end{array}$$

If ASB, then we can interpret B as A plus noise:

$$\begin{array}{c|c} \omega \\ \hline & \\ \end{array} \end{array} \begin{array}{c|c} f_{A}(a|\omega) \end{array} \begin{array}{c|c} a \\ \hline & \\ \end{array} \end{array} \begin{array}{c|c} \psi(b|a) \end{array} \begin{array}{c|c} b \\ \hline & \\ \end{array} \end{array}$$

## Sufficiency => Informativeness

[**Theorem: Sufficiency**  $\implies$  **Informativeness**] If estimator A is sufficient for B, then A is more informative B.

- Prove sufficiency parametrically for:
  - Univariate and bivariate Gaussian likelihoods
  - Lognormal likelihoods
  - Beta likelihoods
- Prove by showing existence of certain matrices
  - Multivariate Gaussian likelihoods

## Using Sufficiency

- Run the simulations
- We found that, for any measure of performance, the error distributions are approximately Gaussian with scale parameter 1
- Determine sufficiency relationships for each measure of performance, each trajectory
- Condition on relevant trajectory parameters
- Determine proportion of trajectories where *ASB*
- ASB does not tell you how much better A is than B, so we also report variance ratios

## Sufficient Comparisons

• Unconditional sufficiency of an adaptive c<sub>min</sub> tracker for a two mode interacting multiple model, constant rate turns

proportion for which  $c_{min} S$  imm

Time	2D Pos	Speed	Heading	Range	Bearing	Rng Rt	X-Rng Rt
$0  \mathrm{sec}$	0.9983	0.9987	0.8674	0.9991	0.9993	0.9980	0.9998
10	0.9970	0.9983	0.8585	0.9987	0.9991	0.9987	1.0000
20	0.9870	0.9813	0.8453	0.7776	0.9957	0.7346	1.0000
30	0.9716	0.9039	0.8149	0.6749	0.9816	0.5371	0.9729
40	0.9119	0.7595	0.6873	0.8312	0.9373	0.6886	0.8088
50	0.8103	0.6226	0.6782	0.9486	0.8451	0.9236	0.5816

variance ratios

 $\sigma^{2}_{imm}$  :  $\sigma^{2}_{c_{min}}$ 

Time	2D Pos	Speed	Heading	Range	Bearing	$\operatorname{Rng} \operatorname{Rt}$	X-Rng Rt
$0  \mathrm{sec}$	2.9614	4.4665	1.3772	5.5660	2.7444	5.9335	5.8156
10	2.8199	4.4796	1.3943	5.1887	2.6207	5.4865	5.6934
20	2.6004	3.2424	1.4630	1.6261	2.5770	1.4042	5.9144
30	2.4479	2.0881	1.4748	1.2522	2.5622	0.9093	4.3019
40	2.0482	1.5423	1.2602	1.6377	1.7810	1.1234	1.9695
50	1.6060	1.2089	1.1962	2.2325	1.3597	1.8474	1.2156

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## Summary

- New technique for comparing trackers fairly
- Comprehensive trajectory sets
- Comprehensive measures of performance
- Sufficient comparisons
  - proportions
  - variance ratios
- Future tracker evaluations