

Particle Filter Speed Up Using a GPU

15 September 2010

High Performance Embedded Computing Workshop

MIT Lincoln Labs

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This work was sponsored by the Office of Naval Research, Code 333, under contract number N00014-05-G-0106/0006. (Any opinions, findings, and conclusions or recommendations expressed in this material do not necessarily reflect the views of ONR.)

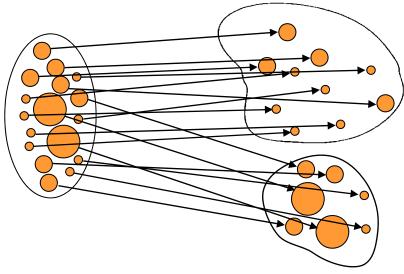


- Things people want to track
 - Physical objects, using radar and sonar (airplanes, ships, fish schools, ...)
 - Time series (financial information, weather statistics, ...)
 - Classification features (spectral lines, ...)
- Tracking is commonly formulated as a probabilistic statespace problem
 - Chapman-Kolmogorov/Fokker-Planck equations for density evolution
 - Bayesian approach to density updating
- There are few closed-form solutions
 - Gaussianity and linearity assumptions lead to the Kalman Filter
- Simple special-case extensions can be made (e.g., Extended Kalman Filter, mixture models)
- General solutions require use of numeric methods
 - Mesh representations
 - Monte Carlo integration (particle filters)



Particle Filters for Tracking in Complex Environments

- Particle Filters are a sequential Monte Carlo methodology for state estimation in which collections of weighted point particles are used to model state probability density functions
- Able to handle complicated non-Gaussian and non-linear problems not easily solvable by Kalman Filters
- Approximate solution to the exact problem, rather than an exact solution to an approximate problem
- Other names
 - Condensation Trackers
 - Bootstrap Filters
 - Survival of the Fittest
- Appropriate for:
 - Applying non-linear constraints
 - Boundaries
 - Kinematic limitations
 - Handling sensor blind spots (exploiting absence of measurements)
 - Tracking "features"
 - Groups of objects





Particle Filters and GPUs: A Marriage Made in Heaven?

- Particle Filters are flexible and powerful, but ...
 - Accuracy increases with the number of particles
 - So does the computation cost
- Graphic Processing Units
 - Commodity items specialized for gaming/imaging applications
 - Highly parallel co-processors
 - Lots of bang for the buck (and watt)
- Particle Filter paradigm appears well-suited to parallel processing
 - Particles can be propagated independently (mostly!)

Particle Filter Algorithm

State PDF Representation Collection of particles $\{(x,w)\}$ **Arbitrary Models** x' = F(x, U) (propagation) z = G(x, V) (measurement) Prediction Step: For each particle: Embarrassingly **Parallel Steps** $\mathbf{x}^{"} = \mathbf{F}(\mathbf{x}, u)$ Correction Step: For each particle: Loop w'' = P(x'|z)wPeriodic resampling of population: Cumulative sum of weights Sorted RNG Possibly Binary search Complicated