Verifying Stochastic Systems

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Motivation

- Robots are often subject to large uncertainties
 - dynamical: wind gusts
 - perceptual: stereo vision
- To maximize performance, want to plan against typical case (99%) rather than worst case (100%)

Lyapunov equations

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 Sufficient condition for stability: nonnegative function V such that



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$$dx(t) = -xdt + dw(t)$$



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- todo: choose and optimize over a family of functions V

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- $E[dV/dt] = p(x)exp(x^TJx)$

• from previous slide: $\mathbb{E}[\dot{V}(x)] = p(x)e^{x^TJx}$

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$$p(x)e^{x^TJx} \le c$$

$$\iff p(x) \le ce^{-x^TJx}$$

$$\iff p(x) \le c(1 - x^TJx)$$

 polynomial condition: optimize with SOS programming

Results: Quadrotor



Results: UAV



Further reading

- http://groups.csail.mit.edu/locomotion/
- LQR-trees tutorial on Friday in the "Integrated Planning and Control" workshop