Consensus theory and Hilbert metric R. Sepulchre University of Liege, Belgium

LCCC workshop January 2010 Consensus and coordination on nonlinear spaces (circle, orthogonal group, SE(2), SE(3), ...)

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Classical linear consensus theory

Linear consensus algorithms are linear time-varying systems

$$x(t+1) = A(t)x(t), \quad x(t) \in \mathbb{R}^n$$

where for each t, A(t) is row stochastic, i.e.

A is nonnegative: $a_{ij} \ge 0$

each row sums to one: A(t)1 = 1

Uniform convergence to $\alpha 1$ ("consensus: $x_i = x_j$ ") is proven under uniform connectivity / irreducibility (*Tsitsiklis*, Jadbabaie et al., Moreau, ...)

Convergence analysis and Lyapunov functions

Tsitsiklis (1986) observed that

$$V(x) = \max_{1 \le i \le n} x_i - \min_{1 \le i \le n} x_i$$

is non increasing along the flow.

Uniform convergence is established by showing the strict decay of over a finite horizon.

It is known that no common quadratic Lyapunov exists in general. (See Olshevsky & Tsitsiklis 08 for a discussion) Let K a closed solid cone in X a Banach space, with partial ordering $\ \preceq$.

A is positive if A maps \mathbf{k} to \mathbf{k}

A is monotone if $x \leq y \Rightarrow Ax \leq Ay$

Theorem (G. Birkhoff, 1957):

Positive linear monotone mappings contract the Hilbert metric in \mathring{K} . The contraction coefficient is $\tanh \frac{1}{4}\Delta(A)$

Note: Perron-Frobenius follows from contraction mapping theorem

Birkhoff Theorem and Hilbert metric in the positive orthant

$$X = \mathbb{R}^n \qquad K = \{(x_1, \ldots, x_n) : x_i \ge 0, 1 \le i \le n\}$$

The Hilbert metric is

$$d(x,y) = \log \frac{\max(x_i/y_i)}{\min(x_i/y_i)}$$

It is a projective metric: $d(\lambda x, \mu y) = d(x, y)$ $\lambda > 0, \mu > 0$

For A>0, the diameter is

$$\Delta(A) = \max\{\log(\frac{a_{ij}a_{pq}}{a_{iq}a_{pj}}) : 1 \le i, j, p, q \le n\}$$

Consequence of Birkhoff result: for nonnegative linear maps that satisfy A(t)1 = 1, the Lyapunov function

 $d(x,1) = \max \log(x_i) - \min \log(x_i)$

is non-increasing along solutions.

The Hilbert distance to consensus is equivalent to Tsitsiklis Lyapunov function in log coordinates.

(and captures the invariance property $d(\lambda x, \mu y) = d(x, y)$).

Remark: both are measures of $co\{x_1, \ldots, x_n\}$ (Moreau's Lyapunov function). Hilbert metric in an arbitrary cone

$$M(x, y) = \inf\{\lambda : x - \lambda y \leq 0\}$$

$$m(x,y) = \sup\{\lambda : x - \lambda y \succeq 0\}$$

$$d(x,y) = \log\{M(x,y)/m(x,y)\}$$

Closely related metric: Thompson metric

$$d_T(x,y) = \log \max\{M(x,y), m^{-1}(x,y)\}$$

Hilbert metric in the SDP cone

$$K = \{ X \in \mathbb{R}^{n \times n} \mid X = X^T \succeq 0 \}$$

$$M(X,Y) = \inf\{\lambda : X - \lambda Y \leq 0\} = \max_{\|v\|=1} \left(\frac{v^T X v}{v^T Y v}\right)$$
$$m(X,Y) = \sup\{\lambda : X - \lambda Y \geq 0\} = \min_{\|v\|=1} \left(\frac{v^T X v}{v^T Y v}\right)$$
$$d(X,Y) = \log\left(\frac{\lambda_{\max}(Y^{-\frac{1}{2}}XY^{-\frac{1}{2}})}{\lambda_{\min}(Y^{-\frac{1}{2}}XY^{-\frac{1}{2}})}\right)$$

Closely related metric

$$d_{\text{Riem}}(X,Y) = \|\log(Y^{-\frac{1}{2}}XY^{-\frac{1}{2}})\|_{F}$$

Generalizations of classical consensus theory

I. Linearity is not essential, only homogeneity (Recent work by Gaubert et al. on generalizations of Perron-Frobenius)

II. Consensus theory generalizes to any cone, e.g. the cone of positive semidefinite matrices.

How to define a consensus iteration over the SDP cone ? What for?

Non-commutative consensus theory

Stochastic maps in non-commutative spaces find applications in

I. Control and estimation of open quantum systems

II. Non-commutative symbolic coding

Stochastic maps: the usual (commutative) case

Probability space:

$$\mathcal{P} = \{ p \in \mathbb{R}^n \mid p_i \ge 0, 1 \le i \le n, \sum_{i=1}^n p_i = 1 \}$$

Stochastic operators map probabilities to probabilities

A:
$$A1 = 1$$
, $a_{ij} \ge 0, 1 \le i, j \le n$

Consensus theory vs existence of a stationary distribution vs graph theoretic interpretation of irreducibility: see Jadbabaie et al.

Stochastic maps: the quantum (non-commutative) case

Probabilities are described by density matrices $\rho = \sum_{i} p_{i} v_{i} v_{i}^{\dagger}$

$$\mathcal{P} = \{ \rho \in \mathbb{C}^{n \times n} \mid \rho = \rho' \succeq 0, \text{ trace}(\rho) = 1 \}$$

Completely positive maps ("quantum channels") map density matrices to density matrices. They are of the form

$$\Phi(\rho) = \sum_{i} L_{i}\rho L_{i}^{\dagger}, \quad \sum_{i} L_{i}^{\dagger}L_{i} = I$$

The dual map $\Psi(\rho) = \sum_{i} L_{i}^{\dagger} \rho L_{i}$ satisfies $\Psi(I) = I$

A non-commutative consensus problem

Repeated interactions of a quantum system give rise to the system

$$\rho(t+1) = \Phi(t)\rho(t)$$

$$\Phi(t)(\rho) = \sum_{i} L_i(t)\rho L_i^{\dagger}(t), \quad \sum_{i} L_i^{\dagger} L_i = I$$

Birkhoff theorem: the Lyapunov function

$$V(\rho) = d(\rho, I) = \log \frac{\lambda_{\max}(\rho)}{\lambda_{\min}(\rho)}$$

is non-increasing along the iterates of the dual system.

Convergence to a stationary density matrix upon irreducibility conditions.

Conclusions

Conic geometries are adapted to consensus theory ... Quadratic Lyapunov functions are'nt ...

Tsitsiklis Lyapunov function is a measure of contraction of the Hilbert metric.

Birkhoff theorem (positive monotone operators contract the Hilbert metric) applies to more general cones, e.g. the SDP cone.

Opens the way to a consensus theory in noncommutative spaces, with a number of possible applications.

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How to bridge the gap between contraction measures and the i/o approach to consensus ?