

Singularly Perturbed Algorithms for Dynamic Average Consensus

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Dynamic Average Consensus

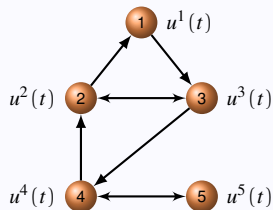
- Autonomous and cooperative agents

$$\dot{x}^i = -c^i, \quad x^i, c^i \in \mathbb{R}$$

- x^i : agreement state
- c^i : driving command

- Design $c^i = f(i, \text{neighbors})$ s.t. $\forall i \in \{1, \dots, N\}$

$$x^i(t) \rightarrow \frac{1}{N} \sum_{j=1}^N u^j(t), \quad t \rightarrow \infty$$



Applications: distributed fusion of dynamic and evolving information

- multi-robot coordination
- distributed tracking
- sensor fusion
- feature-based map merging

Previous literature

- Focus on convergence to consensus
 - Specific initialization conditions: Spanos et al. 05, Zhu and Martinez 10
 - Specific set of inputs: Spanos et al. 05, Olfati-Saber and Shamma 05
 - Track with s.s. error: Olfati-Saber and Shamma 05, Spanos et al. 05, Freeman et al. 06, Zhu and Martinez 10
 - Require knowledge of the dynamics generating inputs: Bai et al. 10
 - Inputs with bounded derivatives: all of them
- No explicit attention to limited control authority
- No explicit attention to rate of convergence of individual agents

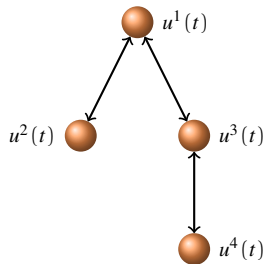
- Dynamic average consensus with pre-specified rate of convergence β :

$$\left| x^i \rightarrow \frac{1}{N} \sum_{j=1}^N u^j(t) \right| \leq k \left| x^i(0) \rightarrow \frac{1}{N} \sum_{j=1}^N u^j(0) \right| e^{-\beta t}$$

- Network of agents with limited control authority; Control over time of arrival
- Dynamic average consensus with pre-specified rate of convergence β^i at each agent:

$$\left| x^i \rightarrow \frac{1}{N} \sum_{j=1}^N u^j(t) \right| \leq k \left| x^i(0) \rightarrow \frac{1}{N} \sum_{j=1}^N u^j(0) \right| e^{-\beta^i t}$$

- Network of agents with different levels of control authority; Control over time of arrival of each agent



Design tool: singular perturbation theory

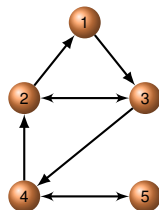
Communication topology: weighted digraph $\mathcal{G}(V, \mathcal{E}, A)$

- Node set: $V = \{1, \dots, N\}$
- Edge set: $\mathcal{E} \subseteq V \times V$
- Weights (for $i, j \in \{1, \dots, N\}$)

$$a_{ij} > 0 \text{ if } (i, j) \in \mathcal{E}, \quad a_{ij} = 0 \text{ if } (i, j) \notin \mathcal{E}$$

- Strongly connected: $i \rightarrow j$ for any i, j
- Weight-balanced:

$$\sum_{j=1}^N a_{ji} = \sum_{j=1}^N a_{ij}, \quad i \in V$$



Goal: Dynamic average consensus with pre-specified rate of convergence β :

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Design methodology

- Simplest dynamics: $x^i \rightarrow \frac{1}{N} \sum_{j=1}^N u^j(t)$ **with rate** β

$$\dot{x}^i = -\beta \left(x^i - \frac{1}{N} \sum_{j=1}^N u^j \right) + \frac{1}{N} \sum_{j=1}^N \dot{u}^j$$

- Requirement:

- fast dynamics to generate $\beta \frac{1}{N} \sum_{j=1}^N u^j + \frac{1}{N} \sum_{j=1}^N \dot{u}^j$ in a **distributed manner** !

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- Requirement:
 - **fast** dynamics to generate $\beta \frac{1}{N} \sum_{j=1}^N u^j + \frac{1}{N} \sum_{j=1}^N \dot{u}^j$ in a **distributed manner** !

Design methodology

- Desired dynamics: $\dot{x}^i = -(\beta x^i - (\beta \frac{1}{N} \sum_{j=1}^N w^j + \frac{1}{N} \sum_{j=1}^N \dot{w}^j))$
- Requirement: **fast** distributed dynamics to generate $\beta \frac{1}{N} \sum_{j=1}^N w^j + \frac{1}{N} \sum_{j=1}^N \dot{w}^j$

- Two-time scale algorithm:

- Initialize at $k = 0$, $x^i(0) \in \mathbb{R}$

- **Fast dynamics:** at each time k , obtain $w^i(k)$ and $\dot{w}^i(k)$. $\forall i \in \{1, \dots, N\}$, run

$$\begin{cases} \dot{z}^i = -(z^i - \beta w^i(k) - \dot{w}^i(k)) - \sum_{i=1}^N a_{ij}(z^i - z^j) - \sum_{i=1}^N a_{ij}(v^i - v^j) \\ \dot{v}^i = \sum_{i=1}^N a_{ji}(z^i - z^j) \end{cases}$$

$$z^i(t, k) \rightarrow \beta \frac{1}{N} \sum_{j=1}^N w^j(k) + \frac{1}{N} \sum_{j=1}^N \dot{w}^j(k), \text{ exponentially as } t \rightarrow \infty, \text{ (Due to [a])}$$

- **Slow dynamics:** $x^i(k+1) = x^i(k) - \Delta t(\beta x^i(k) - z^i(k))$

- $k \leftarrow k + 1$

[a] R. Freeman et al., "Stability and convergence properties of dynamic average consensus estimators," CDC 2006

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An average consensus with pre-specified rate of coverage

- **Fast dynamics:** at each time k , $\forall i \in \{1, \dots, N\}$

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- **Slow dynamics:** $x^i(k+1) = x^i(k) - \Delta t(\beta x^i(k) - z^i(k))$

Innovation

- Combine the fast and slow dynamics in one continuous-time algorithm
- No need to wait for fast dynamics to converge to take steps in the slow dynamics
- Design is based on singularly perturbed systems

$$\begin{cases} \dot{x} = f(t, x, z), & x \in \mathbb{R}^n \\ \epsilon \dot{z} = g(t, x, z), & z \in \mathbb{R}^m, \end{cases} \quad \xrightarrow{\epsilon = 0} \quad \begin{cases} \dot{x} = f(t, x, z), & x \in \mathbb{R}^n \\ 0 = g(t, x, z) \end{cases}$$

Slow dynamics:

$$\begin{aligned} g(t, x, z) = 0 &\Rightarrow z = h(t, x) \\ \dot{x} &= f(t, x, h(t, x)) \end{aligned}$$

Fast dynamics: fixed $(t, x(t))$ and $\tau = t/\epsilon$

$$\frac{dz}{d\tau} = g(t, x, z)$$

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Fast dynamics: fixed $(t, x(t))$ and $\tau = t/\epsilon$

$$\frac{dz}{d\tau} = g(t, x, z)$$

Full dynamics:

$$\begin{cases} \dot{x} = f(t, x, z), \\ \epsilon \dot{z} = g(t, x, z) \end{cases}$$

Solution : $x(t, \epsilon)$

Slow and fast dynamics

$$\begin{cases} \dot{x} = f(t, x, h(t, x)) \\ \frac{dz}{d\tau} = g(t, x, z) \end{cases}$$

Solution : $\bar{x}(t)$

Theorem

For $[t, x, z - h(t, x), \epsilon] \in [0, \infty) \times D_x \times D_y \times [0, \epsilon_0)$

• On any compact subset of $D_x \times D_y$:

- continuous and bounded: $f, g, \partial f|_{\partial x, \partial z, \partial \epsilon}, \partial g|_{\partial x, \partial z, \partial \epsilon, \partial t}$
- bounded partial derivative w.r.t arg: $h(t, x), \partial g(t, x, z, 0) / \partial z$
- $\partial f(t, x, h(t, x), 0) / \partial x$ is Lipschitz in x , uniformly in t ,

• The slow dynamics is exponentially stable

• The fast dynamics is exponentially stable

For any $t_0 \geq 0, \exists \epsilon^*$ s.t. for $0 < \epsilon \leq \epsilon^*$ we have

$$x(t, \epsilon) - \bar{x}(t) \in O(\epsilon), \quad t \in [t_0, \infty)$$

A singularly perturbed dynamic average consensus: pre-specified rate of coverage β

- 1st-Order-Input Dynamic Consensus (FOI-DC)

$$\begin{cases} \epsilon \dot{z}^i = -(z^i + \beta u^i + \dot{u}^i) - \sum_{j=1}^N a_{ij}(z^i - z^j) - \sum_{j=1}^N a_{ij}(v^i - v^j), \\ \epsilon \dot{v}^i = \sum_{j=1}^N a_{ji}(z^i - z^j), \\ \dot{x}^i = -\beta x^i - z^i, \end{cases}$$

- 2nd-Order-Input Dynamic Consensus (SOI-DC)

$$\begin{cases} \epsilon \dot{z}^i = -(z^i + \beta u^i + \dot{u}^i) - \sum_{j=1}^N a_{ij}(z^i - z^j) - \sum_{j=1}^N a_{ij}(v^i - v^j) - \epsilon(\beta \dot{u}^i + \ddot{u}^i), \\ \epsilon \dot{v}^i = \sum_{j=1}^N a_{ji}(z^i - z^j), \\ \dot{x}^i = -\beta x^i - z^i, \end{cases}$$

Convergence analysis: using singular perturbation theory

- FOI-DC

$$\begin{cases} \epsilon \dot{z}^i = -(z^i + \beta u^i + \dot{u}^i) - \sum_{i=1}^N a_{ij}(z^i - z^j) - \sum_{i=1}^N a_{ij}(v^i - v^j), \\ \epsilon \dot{v}^i = \sum_{i=1}^N a_{ji}(z^i - z^j), \\ \dot{x}^i = -\beta x^i - z^i, \end{cases}$$

- SOI-DC

$$\begin{cases} \epsilon \dot{z}^i = -(z^i + \beta u^i + \dot{u}^i) - \sum_{i=1}^N a_{ij}(z^i - z^j) - \sum_{i=1}^N a_{ij}(v^i - v^j) - \epsilon(\beta \dot{u}^i + \ddot{u}^i), \\ \epsilon \dot{v}^i = \sum_{i=1}^N a_{ji}(z^i - z^j), \\ \dot{x}^i = -\beta x^i - z^i, \end{cases}$$

Theorem

- Let \mathcal{G} be a strongly connected and weight-balanced digraph
- FOI-DC: \dot{u}^i and \ddot{u}^i continuous and bounded for $t \geq 0$
- SOI-DC: \dot{u}^i and \ddot{u}^i are continuous and bounded for $t \geq 0$

Then, $\exists \epsilon^* > 0$ s. t., for all $\epsilon \in (0, \epsilon^*]$, $x^i(0), z^i(0), v^i(0) \in \mathbb{R}, \forall i \in \{1, \dots, N\}$

$$|x^i(t, \epsilon) - \frac{1}{N} \sum_{j=1}^N u^j(t)| < O(\epsilon) + |x^i(0) - \frac{1}{N} \sum_{j=1}^N u^j(0)| e^{-\beta t},$$

Sketch of proof:

- Fast dynamics is exponentially stable ($\tau = t/\epsilon$)

$$\begin{cases} dz^i/d\tau = -(z^i + \beta u^i + \dot{u}^i) - \sum_{i=1}^N a_{ij}(z^i - z^j) - \sum_{i=1}^N a_{ij}(v^i - v^j), \\ dv^i/d\tau = \sum_{i=1}^N a_{ji}(z^i - z^j), \end{cases}$$

Setting $\epsilon = 0$: $z^i = \beta \frac{1}{N} \sum_{j=1}^N u^j + \frac{1}{N} \sum_{j=1}^N \dot{u}^j, \forall i \in \{1, \dots, N\}$

- Slow dynamics is exponentially stable

$$\dot{x}^i = -\beta \left(x^i - \frac{1}{N} \sum_{j=1}^N u^j \right) + \frac{1}{N} \sum_{j=1}^N \dot{u}^j \rightarrow |x^i(t) - \frac{1}{N} \sum_{j=1}^N u^j(t)| \leq |x^i(0) - \frac{1}{N} \sum_{j=1}^N u^j(0)| e^{-\beta t}$$

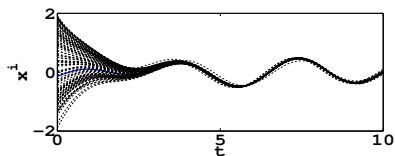
- Lipschitz and continuity conditions are satisfied

$$\therefore \mathbf{x}(t, \epsilon) - \mathbf{x}(t) \in \mathcal{O}(\epsilon)$$

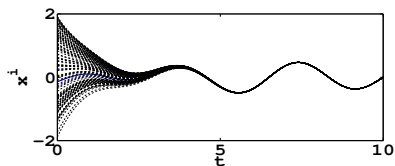
- $N = 100$
- random connected graph
- $u^i(t) = a^i \sin(b^i t + c^i)$
 $a^i \sim \mathcal{U}[-5, 5]$
 $b^i \sim \mathcal{U}[1, 2]$
 $c^i \sim \mathcal{U}[0, \pi/2]$

↓ ϵ : ↓ **error**

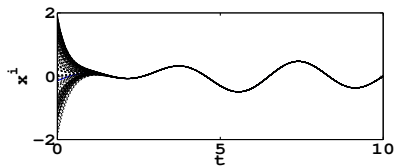
↑ β : **faster convergence**



$\epsilon = 0.01$ and $\beta = 1$



$\epsilon = 0.001$ and $\beta = 1$



$\epsilon = 0.001$ and $\beta = 3$

Lemma

$$\begin{cases} \epsilon \dot{z}^i = -(z^i + \beta u^i + \dot{u}^i) - \sum_{i=1}^N a_{ij}(z^i - z^j) - \sum_{i=1}^N a_{ij}(v^i - v^j) - \epsilon(\beta \dot{u}^i + \ddot{u}^i), \\ \epsilon \dot{v}^i = \sum_{i=1}^N a_{ji}(z^i - z^j), \\ \dot{x}^i = -\beta x^i - z^i, \end{cases}$$

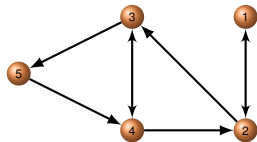
- Let \mathcal{G} be a strongly connected and weight-balanced digraph
- $u^i(t) = u(t) + \bar{u}^i$, $\dot{u}^i = 0$, $\forall i \in \{1, \dots, N\}$

Then, $\forall \epsilon > 0$ and $\beta > 0$ and, $x^i(0), z^i(0), v^i(0) \in \mathbb{R}$, $\forall i \in \{1, \dots, N\}$

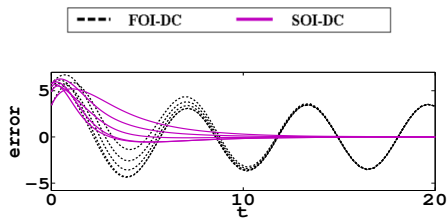
$$x^i(t, \epsilon) \rightarrow \frac{1}{N} \sum_{j=1}^N u^j(t), \quad t \rightarrow \infty$$

Proof is based on Lyapunov approach!

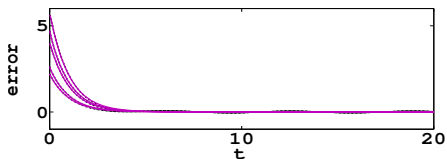
Simulation: SOI-DC for inputs differing by static values



$$\begin{aligned}u^1(t) &= 5 \sin(t) + 1, \\u^2(t) &= 5 \sin(t) - 1, \\u^3(t) &= 5 \sin(t) + 4, \\u^4(t) &= 5 \sin(t) + 5, \\u^5(t) &= 5 \sin(t) + 10.\end{aligned}$$



$\epsilon = 1$ and $\beta = 1$



$\epsilon = 0.01$ and $\beta = 1$

SOI-DC tracks regardless of value of ϵ

Lemma

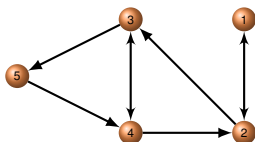
$$\begin{cases} \epsilon \dot{z}^i = -(z^i + u^i) - \sum_{i=1}^N a_{ij}(z^i - z^j) - \sum_{i=1}^N a_{ij}(v^i - v^j), \\ \epsilon \dot{v}^i = \sum_{i=1}^N a_{ji}(z^i - z^j), \\ \epsilon \dot{y}^i = -(y^i + \dot{u}^i) - \sum_{i=1}^N a_{ij}(y^i - y^j) - \sum_{i=1}^N a_{ij}(\mu^i - \mu^j), \\ \epsilon \dot{\mu}^i = \sum_{i=1}^N a_{ji}(y^i - y^j), \\ \dot{x}^i = -\beta^i x^i - \beta^i z^i - y^i, \end{cases}$$

- Let \mathcal{G} be a strongly connected and weight-balanced digraph
- Assume \dot{u}^i and \ddot{u}^i continuous and bounded for $t \geq 0$

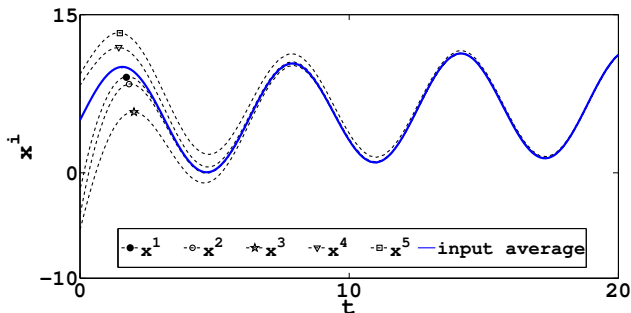
Then, $\forall \epsilon > 0$ and $\beta^i > 0$ and, $x^i(0), z^i(0), v^i(0) \in \mathbb{R}, \forall i \in \{1, \dots, N\}$

$$|x^i(t, \epsilon) - \frac{1}{N} \sum_{j=1}^N u^j(t)| < O(\epsilon) + |x^i(0) - \frac{1}{N} \sum_{j=1}^N u^j(0)| e^{-\beta^i t},$$

Simulation: agents set their own rate of convergence



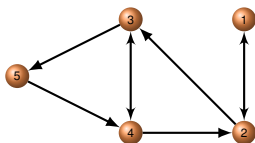
$$\begin{aligned}u^1(t) &= 5 \sin(t) + \frac{10}{t+2} + 1, \\u^2(t) &= 5 \sin(t) + 0.5t - 1, \\u^3(t) &= 5 \sin(t) + \cos(0.5t) + 4, \\u^4(t) &= 5 \sin(t) + \log(t + 1) + 5, \\u^5(t) &= 5 \sin(t) + \text{atan}(t) + 10.\end{aligned}$$



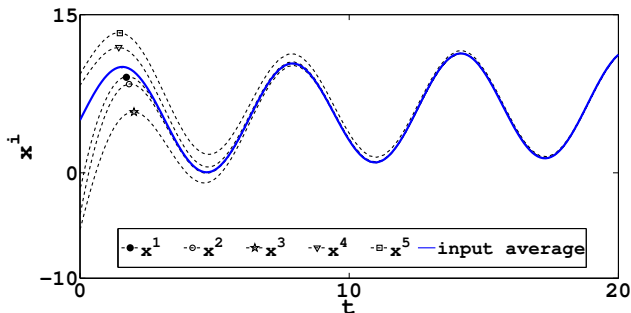
$$\epsilon = 0.01 \text{ and } \beta^1 = 1.2, \beta^2 = 1, \beta^3 = 0.5, \beta^4 = 0.4, \beta^5 = 0.2$$

Control over rate of convergence ~ Control over time of arrival!

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Control over rate of convergence ~ Control over time of arrival!

Summary

- We presented a distributed dynamic average consensus algorithm with pre-specified rate of convergence
- We provided a variation which allows each agent to choose its own rate of convergence
- Our algorithm is suitable for networked systems with limited control authority

Future work

- Quantifying the $O(\epsilon)$
- Rigorous treatment of switching topologies
- Relaxing boundedness and continuity conditions of (\dot{u}^i, \ddot{u}^i)

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