Identifying tractable decentralized control problems on the basis of information structure

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Optimal design of decentralized systems with non-classical information structures

- **Difficulties**: Conceptual and computational
- **Results of this paper**: Consider two general models of decentralized systems and obtain a sequential decomposition for their finite and infinite horizon cases.
- Our models encompass
 - ▷ Standard form (Witsenhausen, 1973)
 - ▷ k-step delay sharing pattern (Walrand and Varaiya, 1978)
 - ▷ Generic team model of Witsenhausen (1988)
- **Main idea**: viewed appropriately, these models are equivalent to POMDPs with functions as control actions
- Numerical solution can be obtained using existing techiques for POMDPs

Model A for two agents



Model A for two agents

• **Plant:** $X_{t+1} = f_t(X_t, U_t^1, U_t^2, W_t)$

• Observations

▷ Common message: ▷ Private message: $Z_t^1 = h_t^1(X_t, N_t^1)$ $Y_t = c_t(X_t, U_{t-1}^1, U_{t-1}^2, Q_t)$ $Z_t^2 = h_t^2(X_t, U_t^1, N_t^2)$

• Agent k

$$\triangleright \quad \text{Control: } \mathbf{U}_{t}^{k} = \mathbf{g}_{t}^{k}(\mathbf{Y}^{t}, \mathbf{Z}_{t}^{k}, \mathbf{M}_{t-1}^{k})$$

- ▷ Memory update: $M_t^k = l_t^k(Y^t, Z_t^k, M_{t-1}^k)$
- **Design** \equiv all control and memory update functions of both agents

• Cost at time t:
$$\rho_t(X_t, U_t^1, U_t^2)$$
. Cost of a design: $E\left\{\sum_{t=1}^T \rho_t(X_t, U_t^1, U_t^2) \middle| Design\right\}$

• **Objective**: Determine an optimal design

Model A for two agents

• Salient features

- Non-classical information structures
- ▷ Sequential system





Consider the model from the point of view of a fictitious common agent

Common Agent

Common agent observes all common messages

• Think of control and memory update functions in two steps

$$\begin{split} \boldsymbol{U}_t^k &= \boldsymbol{g}_t^k(\boldsymbol{Y}^t,\boldsymbol{Z}_t^k,\boldsymbol{M}_{t-1}^k) \\ &= \boldsymbol{\widehat{g}}_t^k(\boldsymbol{Z}_t^k,\boldsymbol{M}_{t-1}^k), \quad \text{where } \boldsymbol{\widehat{g}}_t^k = \boldsymbol{\gamma}_t^k(\boldsymbol{Y}^t) \end{split}$$

Similarly,

$$\begin{split} \boldsymbol{M}_t^k &= \boldsymbol{l}_t^k(\boldsymbol{Y}^t,\boldsymbol{Z}_t^k,\boldsymbol{M}_{t-1}^k) \\ &= \boldsymbol{\hat{l}}_t^k(\boldsymbol{Z}_t^k,\boldsymbol{M}_{t-1}^k), \quad \text{where } \boldsymbol{\hat{l}}_t^k = \boldsymbol{\lambda}_t^k(\boldsymbol{Y}^t) \end{split}$$

Common Agent's viewpoint



Common Agent's viewpoint



Common Agent's viewpoint



• Consider three time steps t^0 , t^1 , and t^2 in time interval t

$$\begin{split} S^0_t &= (X_t, M^1_{t-1}, M^2_{t-1}, U^1_{t-1}, U^2_{t-1}), \qquad O^0_t = Y_t \\ S^1_t &= (X_t, M^1_{t-1}, M^2_{t-1}), \qquad O^1_t = - \\ S^2_t &= (X_t, M^1_t, M^2_{t-1}, U^1_t), \qquad O^2_t = - \end{split}$$

• POMDP with: \triangleright State: S_t^i , \triangleright Obs: O_t^i , \triangleright Control actions: $(\hat{g}_t^k, \hat{l}_t^k)$

From the common agent's viewpoint $\{S_t^0, S_t^1, S_t^2, t = 1, ..., T\}$ is a POMDP (partially observable Markov decision process)

Sequential decomposition

• Information states

$$\pi_{t}^{0} = \Pr\left(S_{t}^{0} \mid Y^{t}, \hat{g}^{1,t-1}, \hat{l}^{1,t-1}, \hat{g}^{2,t-1}, \hat{l}^{t-1}\right)$$
$$\pi_{t}^{1} = \Pr\left(S_{t}^{1} \mid Y^{t}, \hat{g}^{1,t-1}, \hat{l}^{1,t-1}, \hat{g}^{2,t-1}, \hat{l}^{t-1}\right)$$
$$\pi_{t}^{2} = \Pr\left(S_{t}^{2} \mid Y^{t}, \hat{g}^{1,t}, \hat{l}^{1,t}, \hat{g}^{2,t-1}, \hat{l}^{t-1}\right)$$

• Optimality equations

$$\begin{split} V_{T+1}^{0}(\pi_{T+1}^{0}) &\equiv 0, \\ \text{for } t = 1, \dots, T \\ V_{t}^{0}(\pi_{t}^{0}) &= E\left\{V_{t}^{1}(\pi_{t}^{1}) \mid \pi_{t}^{0}\right\}, \\ V_{t}^{1}(\pi_{t}^{1}) &= \min_{\theta_{t}^{1}}\left\{E\left\{V_{t}^{2}(\pi_{t}^{2}) \mid \pi_{t}^{1}, \theta_{t}^{1}\right\}\right\}, \\ V_{t}^{2}(\pi_{t}^{2}) &= \min_{\theta_{t}^{2}}\left\{E\left\{\rho_{t}(X_{t}, U_{t}^{1}, U_{t}^{2}) + V_{t+1}^{0}(\pi_{t+1}^{0}) \mid \pi_{t}^{2}, \theta_{t}^{2}\right\}\right\}, \end{split}$$

where $\boldsymbol{\theta}_t^k = (\boldsymbol{\hat{g}}_t^k, \boldsymbol{\hat{l}}_t^k)$

Models considered in the paper

• Model A

▷ n-agent version of what was presented here

• Model B

- ▷ Model A with no common messages
- Also consider infinite horizon problems

Example – multiaccess broadcast



• MAB Channel

- \triangleright Single user transmits \implies successful transmission
- \triangleright Both users transmit \implies packet collision

• Transmitters

- ▷ Queues with buffer size 1
- Packet held in queue until successful transmission
- Packet arrival is independent Bernoulli process

Example – multiaccess broadcast

• Channel feedback

Both transmitters know if there was no transmission, successful transmission, or a collision

• Policy of transmitters

If packet is available, decide whether or not to transmit based on all past channel feedback

- Objective: Maximize throughput
 - Avoid collisions
 - ▷ Avoid idle

History of multiaccess broadcast

• Hluchyj and Gallager,

"Multiaccess of a slotted channel by finitely many users", NTC 81.

- Considered symmetric arrival rates
- ▷ Restricted attention to "window protocols"
- Ooi and Wornell,

"Decentralized control of multiple access broadcast channels", CDC 96.

- Considered a relaxation of the problem
- ▷ Numerically find optimal performance of the relaxed problem
- ▷ Hluchyj and Gallager's scheme meets this upper bound
- AI Literature
 - ▷ Consider the case of asymmetric arrival rates
 - ▷ Approximate heuristic solutions for small horizons

Multi-access broadcast is equivalent to Model A



- Information state: $\pi_t = \Pr(Z_t^1, Z_t^2 | \text{feedback}), \quad Z_t^k = \{0, 1\}$
- Action Space: $\hat{g}_t^k : \{0, 1\} \rightarrow \{Tx, Don't Tx\}$

Equivalent to a POMDP with finite state and action spaces

Tractability

- Finite horizon problem
 - All system variables are finite valued
- Infinite horizon
 - ▷ All system variables take values in a time-invariant space
 - ▷ The system is **time-homogeneous**

Conclusions

- Sequential decomposition of two general models of decentralized systems
- Equivalent to POMDPs (sometimes to POMDPs with finite state and action spaces)
- Harder to solve than POMDPs due to expansion of state and action spaces.

Thank you