# Output feedback control synthesis for a class of polynomial systems

## Full State-feedback : Primal problem $(P_f)$

supp  $\mu_0(X)$  subject to

$$\mathcal{L}'_{f}\mu + \mathcal{L}'_{g}(\sigma^{+} - \sigma^{-}) + \delta_{0} \times \mu_{0} = \delta_{T} \times \mu_{T}$$

$$[\sigma^{+}]_{k} + [\sigma^{-}]_{k} + [\hat{\sigma}]_{k} = \mu \qquad \forall k \in \{1, \dots, m\}$$

$$\mu_{0} + \hat{\mu}_{0} = \lambda$$

$$[\sigma^{+}]_{k}, [\sigma^{-}]_{k}, [\hat{\sigma}]_{k} \geq 0 \qquad \forall k \in \{1, \dots, m\}$$

$$\mu, \mu_{0}, \hat{\mu}, \mu_{T} \geq 0$$

$$\sup_{p} (\mu_{T}) \subset X_{T}$$

$$\sup_{p} (\mu_{0}) \subset X_{1} \times X_{2}$$

$$\sup_{p} (\hat{\mu}_{0}) \subset X_{1} \times X_{2}$$

$$\sup_{p} (\mu) \subset T \times X_{1} \times X_{2}$$

## Partial State-feedback : Primal problem $(P_p)$

Let  $\Gamma_m = \mathbb{N}_m$  and  $\Lambda = \{\bar{\mu}, \mu_0, \mu_T, \hat{\mu}_0, [\bar{\sigma}^+]_{\Gamma}, [\bar{\sigma}^-]_{\Gamma}, [\hat{\bar{\sigma}}]_{\Gamma}\}$ 

 $\sup_{\Lambda} \mu_0(X)$ 

subject to

$$\mathcal{L}'_f \mu + \mathcal{L}'_g (\sigma^+ - \sigma^-) + \delta_0 \times \mu_0 = \delta_T \times \mu_T$$

$$[\sigma^+]_k + [\sigma^-]_k + [\hat{\sigma}]_k = \mu \qquad \forall k \in \{1, \dots, m\}$$

$$d\nu^* (x_2 | t, x_1) d\bar{\mu} = d\mu$$

$$d\nu^* (x_2 | t, x_1) d[\bar{\sigma}^+]_k = d[\sigma^+]_k \qquad \forall k \in \{1, \dots, m\}$$

$$d\nu^* (x_2 | t, x_1) d[\hat{\sigma}]_k = d[\sigma^-]_k \qquad \forall k \in \{1, \dots, m\}$$

$$d\nu^* (x_2 | t, x_1) d[\hat{\sigma}]_k = d[\hat{\sigma}]_k \qquad \forall k \in \{1, \dots, m\}$$

$$\mu_0 + \hat{\mu}_0 = \lambda$$

$$[\bar{\sigma}^+]_k, [\bar{\sigma}^-]_k, [\hat{\sigma}]_k \geq 0 \qquad \forall k \in \{1, \dots, m\}$$

$$\bar{\mu}, \mu_0, \hat{\mu}, \mu_T \geq 0$$

$$\sup (\mu_T) \subset X_T$$

$$\sup (\hat{\mu}_0) \subset X_1 \times X_2$$

$$\sup (\hat{\mu}_0) \subset X_1 \times X_2$$

$$\sup (\hat{\mu}_0) \subset T \times X_1$$

$$\sup (\bar{\sigma}^\pm) \subset (T \times X_1)^m$$

$$\sup (\hat{\sigma}^\pm) \subset (T \times X_1)^m$$

where  $\mu^*$  is the optimal solution to  $\mathbf{P}_f$  and  $d\mu^* = d\nu^*(x_2|t,x_1)d\pi_{t,x_1}\mu^*(t,x_1)$ .

### Partial State-feedback : Dual problem $(D_p)$

In the dual formulation below, let  $\phi: C_c(T \times X_1 \times X_2) \to C_c(T \times X_1)$  be the operator representation of integrating-out the conditional  $\nu^*(x_2|t,x_1)$ ; i.e.

$$v(t, x_1, x_2) \mapsto \int_{X_2} v(t, x_1, x_2) d\nu^*(x_2 | t, x_1).$$
  $\forall v \in C_c(T \times X_1 \times X_2)$ 

The dual:

 $\inf \langle \lambda, w \rangle$ 

subject to

$$\phi \circ \mathcal{L}_f v + \sum_{k=1}^m \phi \circ [p]_k \le 0$$

$$w - v(0, \cdot) - 1 \ge 0$$

$$v(T, \cdot) \ge 0$$

$$w \ge 0$$

$$\phi \circ [p]_k \ge 0, \quad \phi \circ [p]_k \ge |\phi \circ [\mathcal{L}_g v]_k| \qquad \forall k \in \{1, \dots, m\}$$

where  $w \in C(X_1 \times X_2)$ ,  $v \in C^1(T \times X_1 \times X_2)$ ,  $p_k \in C(T \times X_1 \times X_2)$ ,  $q \in C(T \times X_1 \times X_2)$  and 'o' is used to denote the action of operator  $\phi$  on a function.

**Definition 1** ([3, Definition 1.3.4]). A countably additive measure  $\mu$  on a space  $(X, \Sigma)$  is a probability measure if  $\mu \geq 0$  and  $\mu(X) = 1$ .

**Definition 2.** Suppose  $\mu$  is a measure on the space  $(X, \Sigma)$ . The Jordan-Hahn decomposition of  $\mu$  is unique and given by

$$\mu = \mu^+ - \mu^-$$

where  $\mu^+, \mu^- \geq 0$ . The total variation of  $\mu$  is defined as  $|\mu| = \mu^+ + \mu^-$ . Clearly, if  $\mu \geq 0$ , then  $|\mu| = \mu$ .

**Fact 1** ([4, Theorem 10.4.14]). (Check!!) Let  $\mu$  be an unsigned measure defined on  $(X \times Y, \mathcal{B})$  where  $X \subset \mathbb{R}^n$ ,  $Y \subset \mathbb{R}^m$ ,  $\mathcal{B}$  is the Borel  $\sigma$ -algebra, and  $\pi_y \mu$  be the projection (marginal) onto Y. Then,  $\forall y \in Y$ , there exists an unsigned measure  $\nu(\cdot, y)$  on  $\mathcal{B}$  such that

$$d\mu = d\nu(x_1|x_2) d\pi_{x_2}\mu.$$

Fact 2 ([4, Remark 10.4.4]). (Check!!) If  $\sigma$  is a signed measure on  $(X \times Y, \mathcal{B})$  possessing as regular conditional measures, unsigned measures  $\nu_{\sigma}(\cdot, y)$ ,  $\forall y \in Y$ ,  $|\sigma|$  admits  $\nu_{\sigma}(\cdot, y)$  as its regular conditional measure.

Conversely, if  $\nu(\cdot,y)$  is the regular conditional measure for  $\forall y \in Y$  and it is unsigned, then the regular conditional measures of  $\mu^+$  and  $\mu^-$  are  $\nu(\cdot,y)$ .

**Lemma 1** (Sufficiency). Let  $\mu$  and  $\sigma$  be radon measures on the product space  $X_1 \times X_2$  such that  $\sigma \ll \mu$ , where  $X_1$  and  $X_2$  are Polish spaces. Then each measure can be decomposed as follows

$$d\mu = d\nu_{\mu}(x_2|x_1)d\pi_{x_1}\mu(x_1) d\sigma = d\nu_{\sigma}(x_2|x_1)d\pi_{x_1}\sigma(x_1).$$
(1)

where  $\pi_{x_1}\mu$  is the  $x_1$  marginal of  $\mu$ . If  $d\nu_{\sigma}(x_2|x_1) = d\nu_{\mu}(x_2|x_1)$ , then the Radon-Nikodym derivative  $\frac{d\sigma}{d\mu}$  is not a function of  $x_2$ .

Additionally, if  $\sigma$  and  $\mu$  satisfy the above conditions with  $\mu \geq 0$ ,  $\exists \hat{\sigma} \geq 0$  such that

$$\sigma + \hat{\sigma} = \mu \tag{2}$$

and  $d\hat{\sigma} = d\nu_{\mu}(x_2|x_1) d\hat{\bar{\sigma}}$ . That is, the regular conditional of  $\hat{\sigma}$  given  $x_1$  is identical to that of  $\mu$  and  $\sigma$ .

*Proof.* That each measure can be decomposed as in Eqn. (1) follows from a standard result in measure theory ([4]) and hence we concentrate on the Radon-Nikodym derivative. Since  $\sigma \ll \mu$ , from the definition of marginals,  $\pi_{x_1}\sigma \ll \pi_{x_1}\mu$ ; let  $d\pi_{x_1}\sigma = \phi(x_1) d\pi_{x_1}\mu$ .

Consider an arbitrary test function  $v(x_1, x_2)$ ; then the following equalities holds by definition.

$$\begin{split} g(x_1) &:= \int_{X_2} v(x_1, x_2) \, d\nu_\sigma(x_2|x_1) = \int_{X_2} v(x_1, x_2) \, d\nu_\mu(x_2|x_1) \\ \int_{X_1 \times X_2} v(x_1, x_2) \, d\sigma &= \int_{X_1} g(x_1) \, d\pi_{x_1} \sigma(x_1) \\ &= \int_{X_1 \times X_2} \phi(x_1) \, g(x_1) \, d\pi_{x_1} \mu(x_1) \\ &= \int_{X_1 \times X_2} \phi(x_1) \, v(x_1, x_2) \, d\mu \end{split}$$

Thus  $d\sigma = \phi(x_1) d\mu$  and the Radon-Nikodym derivative is not a function of  $x_2$ .

The existence of  $\hat{\sigma} \geq 0$  such that Eqn. (2) holds since  $\sigma \ll \mu$ . Additionally, as  $\sigma$  and  $\mu$  share the same conditional with respect to  $x_1$ ,  $\sigma$  admits the following decomposition.

$$d\hat{\sigma} = d\nu(x_2|x_1) d(\pi_{x_1}\mu - \pi_{x_1}\sigma)(x_1) = d\nu(x_2|x_1) d\hat{\sigma}(x_1)$$

**Lemma 2.** The solution to the primal problem characterizes the BRS when using partial state-feedback control.

*Proof.* In this proof, we consider the special case of single input control; a more general version follows naturally. For convenience, denote by  $\mathcal{U}_s$  and  $\mathcal{U}_p$  respectively, the set of all admissible state-feedback and partial-state-feedback control laws; by definition,  $\mathcal{U}_p \subseteq \mathcal{U}_s$ . Similarly, let  $\mathcal{X}_s$  and  $\mathcal{X}_p$  represent the collection of admissible state trajectories;  $\mathcal{X}_p \subseteq \mathcal{X}_s$ .

By definition, for any initial condition  $x_0 \in X_0^o$ , there exists a partial state-feedback control law  $u(t, x_1)$  such that the resulting state trajectory is admissible. Thus, for any initial measure  $\mu_0$ , there exist measures  $\mu$ ,  $\mu_T$ ,  $\sigma$  which satisfy the following constraints

$$\delta_T \times \mu_T = \delta_0 \times \mu_0 + \mathcal{L}_f' \mu + \mathcal{L}_q' \sigma$$

such that  $\mu_0$ ,  $\mu_T$ ,  $\mu \geq 0$ , supp  $(\mu_0) \subset \mathcal{X}$ , supp  $(\mu_T) \subset X_T$  and supp  $(\mu) \subset T \times X_1 \times X_2$ . This assertion follows from applying the result in [1, Lemma 1] to the state-feedback problem posed in [2] for any feasible trajectory and reviewing the relation between  $\mathcal{U}_s$  and  $\mathcal{U}_p$ , and  $\mathcal{X}_s$  and  $\mathcal{X}_p$ .

Since the Euclidean space is Polish and hence Souslin, and separable, it follows that  $\mu$  can be decomposed as follows [4, Corollary 10.4.13]

$$d\mu = d\nu(x_2|t, x_1) d\pi_{t,x_1} \mu(t, x_1).$$

Now, since the occupation measure can be interpreted as the time spent in the region of the product space, naturally, the  $(t, x_1)$  marginal  $(\pi_{t,x_1}\mu)$ , is the total time spent by all feasible state trajectories in a slice of the product space; and  $\nu(x_2|t,x_1)$  can be interpreted as measure of how well the trajectories are distributed along the  $x_2$  direction at every time instant (??). Naturally, if  $\nu^*(A|t,x_1) = 0$ , then  $\nu(A|t,x_1) = 0 \ \forall (t,x_1) \in T \times X_1$   $(\because \mathcal{X}_p \subseteq \mathcal{X}_s)$ ; and hence  $\nu(\cdot|t,x_1) \ll \nu^*(\cdot|t,x_1)$ . Thus, using [5, Theorem 58],  $\exists \phi(t,x_1)$ , a  $\pi_{t,x_1}\mu$  measurable function such that

$$d\mu = \phi(t, x_1) d\nu^*(x_2|t, x_1) d\pi_{t, x_1} \mu(t, x_1),$$
  
=  $d\nu^*(x_2|t, x_1) d\bar{\mu}(t, x_1),$ 

where the final equality is obtained by appropriately defining  $\bar{\mu}$ .

Since  $\sigma$  and  $\mu$  are solutions to the Liouville equation with R-N derivative  $u = u(t, x_1)$ , the following relations hold

$$d\sigma = u \, d\mu,$$
  
=  $u d\nu^*(x_2|t, x_1) \, d\bar{\mu}(t, x_1),$   
=  $d\nu^*(x_2|t, x_1) \, d\bar{\sigma}(t, x_1),$ 

with  $d\bar{\sigma} = u d\bar{\mu}$ .

Define  $\sigma^+, \sigma^- \geq 0$  such that the following holds (Jordan decomposition)

$$\sigma = \sigma^+ - \sigma^-$$

The regular conditional measures of  $\sigma^+$  and  $\sigma^-$ ,  $\forall y \in Y$  are both  $\nu^*(\cdot|t, x_1)$  (Facts 1 & 2). Consequently,  $\sigma^{\pm}$  can be decomposed as follows

$$d\sigma^{+} = d\nu^{*}(x_{2}|t, x_{1}) d\bar{\sigma}^{+}(t, x_{1})$$
$$d\sigma^{-} = d\nu^{*}(x_{2}|t, x_{1}) d\bar{\sigma}^{-}(t, x_{1}).$$

Finally, to see that  $\hat{\sigma}$  shares the same conditional as  $\mu$ , use Fact 2 and Lemma 1.

Thus, it follows that for every initial condition in the BRS, there exists a feasible solution to  $\mathbf{P}_p$  and hence the optimization problem  $\mathbf{P}_p$  is stronger that any other any problem that exactly identifies the BRS. Thus if  $q^*$  is the optimal value of the cost, then  $q^* \geq \lambda(X_0^o)$ .

Next, we show that  $q^* \leq \lambda(X_0^o)$  by contradiction. Let  $(\mu_0, \mu_T, \mu, \sigma^+, \sigma^-, \hat{\sigma})$  be a feasible solution to  $\mathbf{P}_p$  and suppose  $\lambda(A := \text{supp } (\mu_0) \backslash X_0^o) \neq 0$ . By definition,  $\exists u = u(t, x_1)$  such that  $d\sigma = u d\mu$ ; re-define the dynamics of the system by subsuming this control law into the drift vector field; i.e.

$$\dot{x} = \bar{f}(t, x_1, x_2) = f(t, x_1, x_2) + u(t, x_1)g(t, x_1, x_2).$$

It is evident that the tuple  $(\mu_0, \mu, \mu_T)$  is a feasible solution to the optimization problem

$$\delta_0 \times \mu_0 = \delta_T \times \mu_T + \mathcal{L}'_{\bar{r}}\mu.$$

Using [1, Lemma 3], it then follows that there exists a family of admissible trajectories emanating from supp  $(\mu_0)$  and terminating in supp  $(\mu_T)$ . This is a contradiction since trajectories starting from supp  $(\mu) \setminus X_0^o$  cannot be admissible; i.e.  $\lambda(X_0^o) \geq \lambda(\text{supp }(\mu_0))$ . Thus,  $\lambda(X_0^o) = \lambda(\text{supp }(\mu_0))$ .

The next Lemma is a variation of [1, Lemma 2] in that it establishes that the BRS can be identified as the interior of the super-level set  $\{(x_1, x_2) \mid w(x_1, x_2) \geq 1\}$ .

**Lemma 3.** Let  $(v, w, p_{1,...,m})$  be a tuple of feasible solutions to the dual  $\mathbf{D}_p$ . Then  $v(0,\cdot,\cdot) \geq 0$  and  $w \geq 1$  on  $X_0 (:= supp (\mu_0))$ .

*Proof.* By the fundamental theorem of calculus and the constraints of  $\mathbf{D}_p$ ,

$$\begin{split} 0 &\leq v(T,x(T)) = v(0,x(0)) + \int_{0}^{T} \mathcal{L}_{f}v + \mathcal{L}_{g}v \, u(t) \, dt \\ &= v(0,x(0)) + \int_{T \times X_{1} \times X_{2}} \left( \mathcal{L}_{f}v + \mathcal{L}_{g}v \, u(t,x_{1},x_{2}) \right) d\mu \\ &= v(0,x(0)) + \int_{T \times X_{1}} \left( \phi \circ \mathcal{L}_{f}v + \phi \circ \left( \mathcal{L}_{g}v \, u(t,x_{1},x_{2}) \right) \right) d\bar{\mu} \\ &= v(0,x(0)) + \int_{T \times X_{1}} \left( \phi \circ \mathcal{L}_{f}v + \sum_{i=1}^{m} \phi \circ [p]_{i} + \phi \circ \left( \mathcal{L}_{g}v \, u(t,x) \right) - \sum_{i=1}^{m} \phi \circ [p]_{i} \right) d\bar{\mu} \\ &\leq v(0,x(0)) \leq w(x(0)) - 1 \end{split}$$

where  $x(t) = (x_1(t), x_2(t)).$ 

#### Notes / todo

- 1. Check to ensure that all text in red make sense.
- 2. Why is it okay to optimize over just  $\bar{\mu} \in \mathcal{M}(T \times X_1)$ ? Use result in [4, Theorem 10.7.2].
- 3. Using the regular conditional from an unsigned measure and a positive 'marginal', we can construct another unsigned measure?
- 4. Evaluating output-feedback controllability using inner approximations
- 5. Change notations to ensure there is no confusions between the final time T and the interval T = [0, T].
- 6. Having changed the primal problem formulation, need to show that the  $\hat{\sigma}$  has the same conditional as  $\mu$ . (Lemma 2)
- 7. Check to see if the assumption of controllability is required and the kind of dynamical systems can be entertained [1, Assumption 2].

## References

- [1] Henrion, D.; Korda, M., "Convex Computation of the Region of Attraction of Polynomial Control Systems," Automatic Control, IEEE Transactions on , vol.59, no.2, pp.297,312, Feb. 2014 doi: 10.1109/TAC.2013.2283095
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