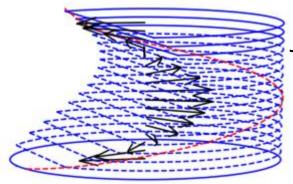
# **Time-Varying SDPs**

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Joint work with Amir Ali Ahmadi







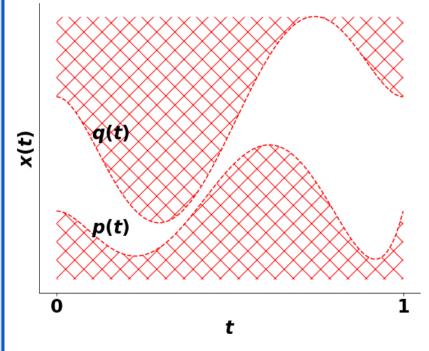
### Toy example

#### **Problem:**

Find a path  $x:[0,1]\to\mathbb{R}$ 

with minimum length 
$$\int_0^1 \sqrt{1+x'(t)^2} \; \mathrm{d}t$$

Such that





### **Framework**

**SDP**: Find  $x \in \mathbb{R}^n$  that minimizes  $\langle c, x \rangle$  such that

$$A_0 + \sum A_i x_i \succeq 0$$

$$A_i \in \mathbb{R}^{m \times m}$$

**TV-SDP**: Find  $x:[0,1]\to\mathbb{R}^n$  that minimizes  $\int_0^1\langle c(t),x(t)\rangle dt$  such that  $\forall t\in[0,1]$ 

$$Fx(t) := A_0(t) + \sum A_i(t)t)(t)(t) \sum 0 \int_0^t D_i(t,s)x_i(s) ds \succeq 0$$

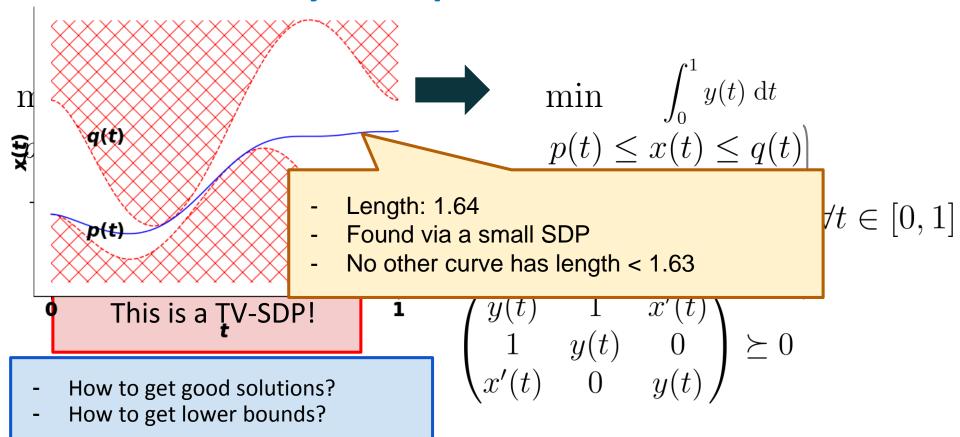
F is linear operator that takes a vectorvalued function x(t) and outputs a matrix-valued function Fx(t)

Data: c,  $A_i$ ,  $D_i$  polynomials. Polynomials are general enough.





### Toy example - continued



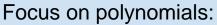




### **Plan**

**Primal approach:** Getting upper bounds.

Any feasible solution will do!



- Smooth
- Tractable



When are polynomial solutions optimal?

**Optimal value of TV-SDP** 



When are these lower bounds tight?

**Dual approach:** Getting lower bounds.

Finite dimensional outer-approxmiation!



### Primal approach

**Our Constraint:** 

 $Fx(t) \succeq 0 \quad \forall t \in [0,1]$ 

Idea: Restrict the search space to polynomials of a given degree.

$$x(t) \in \mathbb{R}^n_{\hat{d}}[t] \implies Fx(t) \in \mathbb{R}^{m \times m}_d[t]$$

$$p(x_1,\ldots,x_n)$$
 sos if  $p(x)=\sum q_i(x)^2$ 

**LP** case studied in [Bampou&Kuhn 12]

SOS polynomials:  $p(x_1,\ldots,x_n)$  sos if  $p(x)=\sum q_i(x)^2$  | Positivstellensatz [Dette&Studden 02] For  $X(t)\in\mathbb{R}_d^{m\times m}[t]$   $X(t)\succeq 0 \ \forall t\in[0,1]$  if and only if  $y^TX(t)y=p_0(t,y)+tp_1(t,y)$ 

$$y^{T}X(t)y = p_{0}(t,y) + tp_{1}(t,y) + (1-t)p_{2}(t,y) + t(1-t)p_{3}(t,y)$$

with  $p_0, \dots, p_3$  SOS of deg. 2 in y and d in t.

Takeaway: (Small) SDP can find the best polynomial solution to a TV-SDP

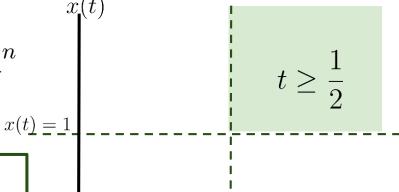




## **Optimality of polynomial solutions (1/2)**

### **Strategy:**

- Take any feasible solution  $x:[0,1] \to \mathbb{R}^n$
- Approximate it with a polynomial



### What can go wrong?

A "discontinuous" TV-SDP

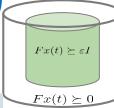
$$\left| (t - \frac{1}{2})x(t) \ge 0, (t - \frac{1}{2})(x(t) - 1) \ge 0 \right|$$





Optimality of polynomial solutions (2/2)

**Definition (Strict feasiblity)**  $\exists \varepsilon > 0 \ \exists x : [0,1] \to \mathbb{R}^n \ Fx(t) \succeq \varepsilon I$ 

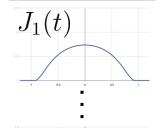


#### Mollification

Continuous func.

Weierstrass

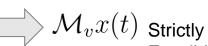
x(t) strictly feasible

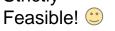


 $J_v(t)$ 











Polynomial Solution

- $1)Fx(t) \succeq 0 \implies \mathcal{M}_v Fx(t) \succeq 0$
- $2)\sup_{t} \|\mathcal{M}_{v}Fx(t) F\mathcal{M}_{v}x(t)\| \underset{v \to \infty}{\longrightarrow} 0$

Theorem (Ahmadi, BEK)

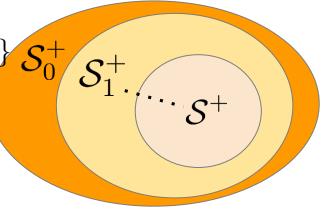
Strict feasibility  $\implies$  Optimality of polynomials



## Dual approach (1/3)

Our constraint 
$$Fx \in \mathcal{S}^+:=\{X \mid X(t) \succeq 0 \ \forall t \in [0,1]\}$$

Inner product  $\langle X, Y \rangle := \int_0^1 Tr(X(t)Y(t)) dt$  $\mathcal{S}^+$  is self dual  $X \in \mathcal{S}^+ \iff \langle X, Y \rangle \geq 0 \ \forall Y \in \mathcal{S}^+$ 



#### Idea:

Restrict Y to be a polynomial of degree < d

$$\mathcal{S}_d^+ := \{ X \mid \langle X, Y \rangle \ge 0 \ \forall Y \in \mathcal{S}^+ \cap \mathbb{R}_d^{m \times m}[t] \}$$



## Dual approach (2/3)

$$\mathcal{S}^{+} := \{X \mid \langle X, Y \rangle \geq 0 \ \forall Y \in \mathcal{S}^{+}\} \quad \text{Initial feasible set}$$
 
$$\mathcal{S}^{+}_{d} := \{X \mid \langle X, Y \rangle \geq 0 \ \forall Y \in \mathcal{S}^{+} \cap \mathbb{R}^{m \times m}_{d}[t]\} \quad \text{Outer-approximation}$$
 
$$Fx \in \mathcal{S}^{+}_{d} \iff \forall Y \in \mathcal{S}^{+} \cap \mathbb{R}^{m \times m}_{d}[t] \qquad \langle Fx, Y \rangle \geq 0$$
 
$$\langle x, F^{*}Y \rangle \geq 0$$
 
$$= \int_{0}^{1} \langle x(t), F^{*}Y(t) \rangle \mathrm{d}t \qquad \text{Depends only on the first moments of } x$$
 
$$\iff x \in \text{dual of } F^{*}(\mathcal{S}^{+} \cap \mathbb{R}^{m \times m}_{d}[t]) \qquad \text{Semidefinite}$$

**Takeaway:** Lower bound at level d can be obtained via SDP



representable

Dual Approach (3/3)

Level	d		$\infty$	
Optimal value	$v^d$	2	$v^{\infty}$	
Optimal solution	$x^d$	<b>_?</b> ⇒	$x^{\infty}$	
Constraint	$\langle Fx, Y \rangle \ge 0$ $\forall Y \in \mathcal{S}^+ \cap \mathbb{R}_d^{m \times m}$	$\langle F \rangle$	$\forall X, Y \geq 0$ $\forall Y \in S^+$	$Fx(t) \succeq 0$

#### Weak convergence

$$\langle x^d, \cdot \rangle \to \langle x^\infty, \cdot \rangle$$

#### **Boundedness**

$$\forall p \in \mathbb{R}_d^n[t]$$
  $p(t) \ge 0$   
 $|\langle x^d, p \rangle| \le M\langle 1, p \rangle$ 

### Theorem (Ahmadi, BEK)

Boundedness of TV-SDP \_\_\_\_\_\_ Lower bounds tight

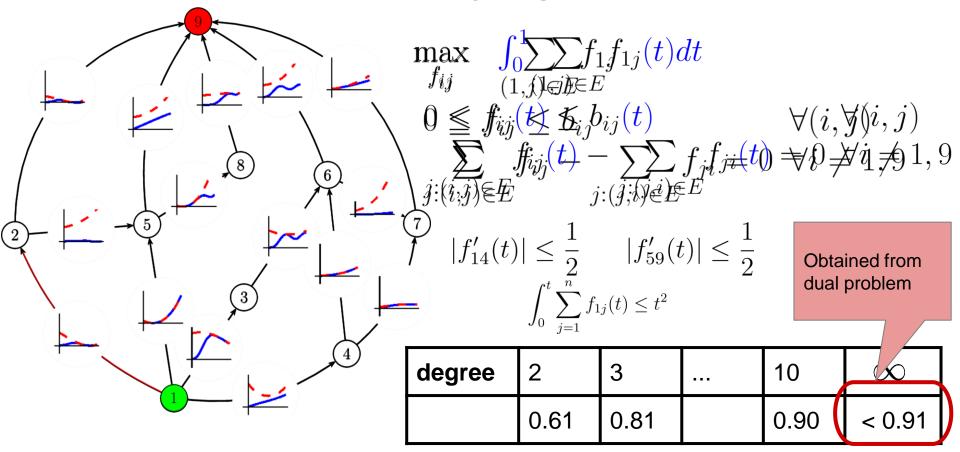
#### **Compactness** [Lasserre]

A bounded sequence has a weakly convergent subsequence





### **TV-maxflow**







### **Pareto Curve approximation**

 $\min \langle c_1, x \rangle$  and  $\langle c_2, x \rangle$  s.t.  $x \in \mathcal{C}$ 

#### **Pareto Curve:**

 $\{(\langle c_1, x(t) \rangle, \langle c_2, x(t) \rangle) \mid t \in [0, 1]\}$   $x(t) := \underset{\langle c_2, x(t) \rangle}{\operatorname{arg min}} \langle c_1, x(t) \rangle$   $\langle c_2, x(t) \rangle \leq t \quad x(t) \in \mathcal{C}$ 

Idea:  $\min \int_0^{\cdot} \langle c_1, x(t) \rangle dt$   $\langle c_2, x(t) \rangle \leq t \quad x(t) \in \mathcal{C}$ 

**Takeaway:** This is a TV-SDP!

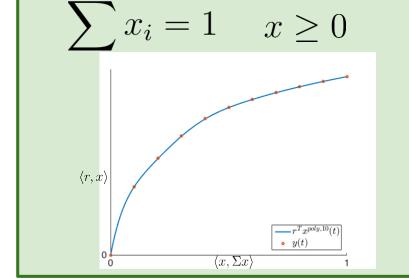
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[Gorissen, Hertog '12]

[Magron, Henrion, Lasserre '14]

### **Markowitz Portfolio Theory**

 $\max \langle r, x \rangle \quad \min \langle x, \Sigma x \rangle$ 





#### Thanks!

# Want to know more? bachirelkhadir.com

