

Quantifier Elimination and Craig Interpolation: The Quantitative Way

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$$\text{QE}(\exists y: x \leq y \wedge y \leq z) = x \leq z$$

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- ▶ Template-based invariant synthesis [Colón et al. '03]:

$$\exists \text{ template instance: } \forall \text{ states: invariant condition holds}$$

- ▶ Symbolic pre-image computations [Komuravelli et al. '14]:

$$\exists \text{ succ: } \text{Trans}(\text{pred}, \text{succ}) \wedge \text{Goal}(\text{succ})$$

Craig Interpolation [Craig '57]

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$$x = y_1 \wedge y_1 \leq y \quad \models \quad \underbrace{x \leq y}_{\text{concise explanation of entailment}} \quad \models \quad z = x \longrightarrow z \leq y$$

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Application: Invariant synthesis from finite loop unrollings [McMillan '03, '06]

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loop unrolling \models potential loop invariant \models specification
concrete explanation of entailment

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“assertions” evaluate to **numbers** instead of truth values

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3. Weighted programming [OOPSLA '22]

What is the competitive ratio of an online algorithm?

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Contribution 1: Quantifier Elimination for Piecewise Linear Quantities

Given $f \in \text{LinQuant}$, compute *quantifier-free* $\text{QE}(f) \in \text{LinQuant}$ with $\text{QE}(f) \equiv f$.

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- ▶ **Soundness and termination** for discontinuous or $\{\infty, -\infty\}$ -valued quantities
- ▶ **Upper space complexity bounds**

Contribution 2: Quantitative Craig Interpolation for Piecewise Linear Quantities

Let $f, f' \in \text{LinQuant}$ with $f \models f'$. There **exists effectively constructible** $g \in \text{LinQuant}$ with

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Piecewise Linear Quantities

Quantitative Quantifier Elimination

Quantitative Craig Interpolation

Summary & Future Work

Piecewise Linear Quantities

The quantitative analogue of Linear Rational Arithmetic

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Syntax: *Piecewise linear quantities* in LinQuant are expressions f of the form

$$Q_1x_1 : \dots : Q_kx_k : [\varphi_1] \cdot a_1 + \dots + [\varphi_n] \cdot a_n ,$$

where

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Semantics: Let $\Sigma = \{\sigma \mid \sigma: \text{Vars} \rightarrow \mathbb{Q}\}$. Then, for quantifier-free $f \in \text{LinQuant}$ and $\sigma \in \Sigma$,

$$\llbracket f \rrbracket(\sigma) = \sum_{i=1}^n \begin{cases} \llbracket a_i \rrbracket(\sigma) & \text{if } \sigma \models \varphi_i \\ 0 & \text{if } \sigma \not\models \varphi_i . \end{cases}$$

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Quantitative Quantifier Elimination

Goal:

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First Observation: We have

$$\begin{aligned} & \text{QE}(Q_1x_1 : \dots : Q_kx_k : [\varphi_1] \cdot a_1 + \dots + [\varphi_n] \cdot a_n) \\ = & \text{QE}(\dots \text{QE}(Q_kx_k : [\varphi_1] \cdot a_1 + \dots + [\varphi_n] \cdot a_n)) \end{aligned}$$

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\implies **Focus on eliminating a single quantifier.**

New Goal:

Given $f = Qx: [\varphi_1] \cdot a_1 + \dots + [\varphi_n] \cdot a_n$, compute $\text{QE}(f) \in \text{LinQuant}$ with

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Our Approach: Divide-And-Conquer in 3 Stages

Stage 1

The Guarded Normal Form

Definition (Guarded Normal Form)

$f = \mathfrak{Q}x: [\varphi_1] \cdot a_1 + \dots + [\varphi_n] \cdot a_n$ is in *Guarded Normal Form*, if:

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$$\exists x: [x \leq y \vee x \leq z_1] \cdot x + [x \leq z_1] \cdot y$$

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$$\begin{aligned} & \mathfrak{Q}x: [x \leq y \vee x \leq z_1] \cdot x + [x \leq z_1] \cdot y \\ \equiv & \mathfrak{Q}x: [(x \leq y \vee x \leq z_1) \wedge x \leq z_1] \cdot (x + y) + [(x \leq y \vee x \leq z_1) \wedge \neg(x \leq z_1)] \cdot x \\ & + [\neg(x \leq y \vee x \leq z_1) \wedge x \leq z_1] \cdot y + [\neg(x \leq y \vee x \leq z_1) \wedge \neg(x \leq z_1)] \cdot 0 \end{aligned}$$

Theorem

If $f = \mathcal{Z}x: [\varphi_1] \cdot a_1 + \dots + [\varphi_n] \cdot a_n$ is in Guarded Normal Form, then

$$\text{QE}(f) = \text{MAX}(\text{QE}(\mathcal{Z}x: [\varphi_i] \cdot a_i + [\neg\varphi_i] \cdot (-\infty)) \mid i \in \{1, \dots, n\}) .$$

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\implies Focus on $\text{QE}(\mathcal{Q}x: [\varphi_i] \cdot a_i + [\neg\varphi_i] \cdot (-\infty))$.

Stage 2

Exploiting the Disjunctive Normal Form

Theorem

For $f = \mathcal{Q}x: [\varphi] \cdot a + [\neg\varphi] \cdot (-\infty)$, where

$$\varphi = \bigvee_{i=1}^n D_i$$

is in DNF, we have

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\implies **Focus on** $\text{QE}(\mathcal{Q}x: [D_i] \cdot a + [\neg D_i] \cdot (-\infty))$,

where D_i is a conjunction of linear inequalities.

Stage 3

Focus on $\text{QE}(\exists x: [D_i] \cdot a_i + [\neg D_i] \cdot (-\infty))$,

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Let $\sigma \in \Sigma$ be a variable valuation.

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Classic Fourier-Motzkin Elimination: $\exists x: D \equiv z < y_1 \wedge z \leq -y_2$

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2. If $\sigma \models \exists x: D$, **maximize** $2x$.

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$$\text{QE}(f) = \underbrace{[\neg(z < y_1 \wedge z \leq -y_2)]}_{\equiv \neg \exists x: D} \cdot (-\infty)$$

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2. If $\sigma \models \exists x: D$, maximize $2x$. **Case distinction on upper bounds on x .**

$$\begin{aligned} \text{QE}(f) &= \underbrace{[\neg(z < y_1 \wedge z \leq -y_2)]}_{\equiv \neg \exists x: D} \cdot (-\infty) \\ &+ \underbrace{[z < y_1 \wedge z \leq -y_2]}_{\equiv \exists x: D} \cdot \underbrace{[y_1 \leq -y_2]}_{y_1 \text{ is sup}} \cdot 2y_1 \end{aligned}$$

$$f = \exists x: \underbrace{[x < y_1 \wedge x \leq -y_2 \wedge x \geq z]}_{=D} \cdot \underbrace{2x}_{=a} + [\neg D] \cdot (-\infty)$$

Let $\sigma \in \Sigma$ be a variable valuation.

1. If $\sigma \not\models \exists x: D$, evaluate to $-\infty$.

Classic Fourier-Motzkin Elimination: $\exists x: D \equiv z < y_1 \wedge z \leq -y_2$

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If $f \in \text{LinQuant}$ contains a single quantifier and

- $|f|_{\rightarrow} = n$ is the number of summands in f , and*
- $|f|_{\downarrow} = m$ is the maximal size of the Boolean expressions in f ,*

then

$$|\text{QE}(f)|_{\rightarrow} \leq n \cdot 2^m \cdot (m + 2)^{n \cdot 2^m} \quad \text{and} \quad |\text{QE}(f)|_{\downarrow} \leq n \cdot 2^m \cdot \left(\left(\frac{m+2}{2} \right)^2 + m + 1 \right).$$

Quantitative Craig Interpolation

Definition

Given $f, f' \in \text{LinQuant}$ with $f \models f'$, we call $g \in \text{LinQuant}$ a *Craig interpolant* of (f, f') , if

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Theorem

Whenever $f \models f'$ there exists an *effectively constructible* Craig interpolant g of (f, f') .

Proof.

Using quantitative quantifier elimination. □

Quantitative Craig Interpolation

$$f \models \underbrace{\text{QE}(\underbrace{\exists x_1 \dots \exists x_n : f}_{= \text{FV}(f) \setminus \text{FV}(f')})}_{\text{strongest Craig interpolant}} \models \underbrace{\text{QE}(\underbrace{\forall y_1 \dots \forall y_m : f'}_{= \text{FV}(f') \setminus \text{FV}(f)})}_{\text{weakest Craig interpolant}} \models f'$$

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2. Upper space complexity bound
3. First Craig interpolation theorem for LinQuant

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Thank you!

Maybe put construction of max