
Satisfiability Modulo Transcendental Functions via Incremental Linearization

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■ Main idea

- Abstract **transcendental functions** with uninterpreted functions
- Incrementally add upper- and lower- bound linear lemmas
 - Tangent and secant lines
 - Added to refine spurious models

■ Challenges

- **Irrational values**: transcendental functions give irrational outputs to (most) rational inputs
- Linearization requires **calculation of slope** at arbitrary point, which is not straightforward
- Handling **periodicity** (of trigonometric functions)
- Detecting **SAT**

Some Math Background

- **Transcendental function** $f(x)$: doesn't satisfy a polynomial equation

- We assume to be continuous and (n-times) differentiable

- **Tangent line** to $f(x)$ at point a : $T_{f,a}(x) \stackrel{\text{def}}{=} f(a) + \frac{d}{dx} f(a) * (x - a)$

- **Secant line** to $f(x)$ between a and b :

$$S_{f,a,b}(x) \stackrel{\text{def}}{=} \frac{f(a)-f(b)}{a-b} * (x - a) + f(a)$$

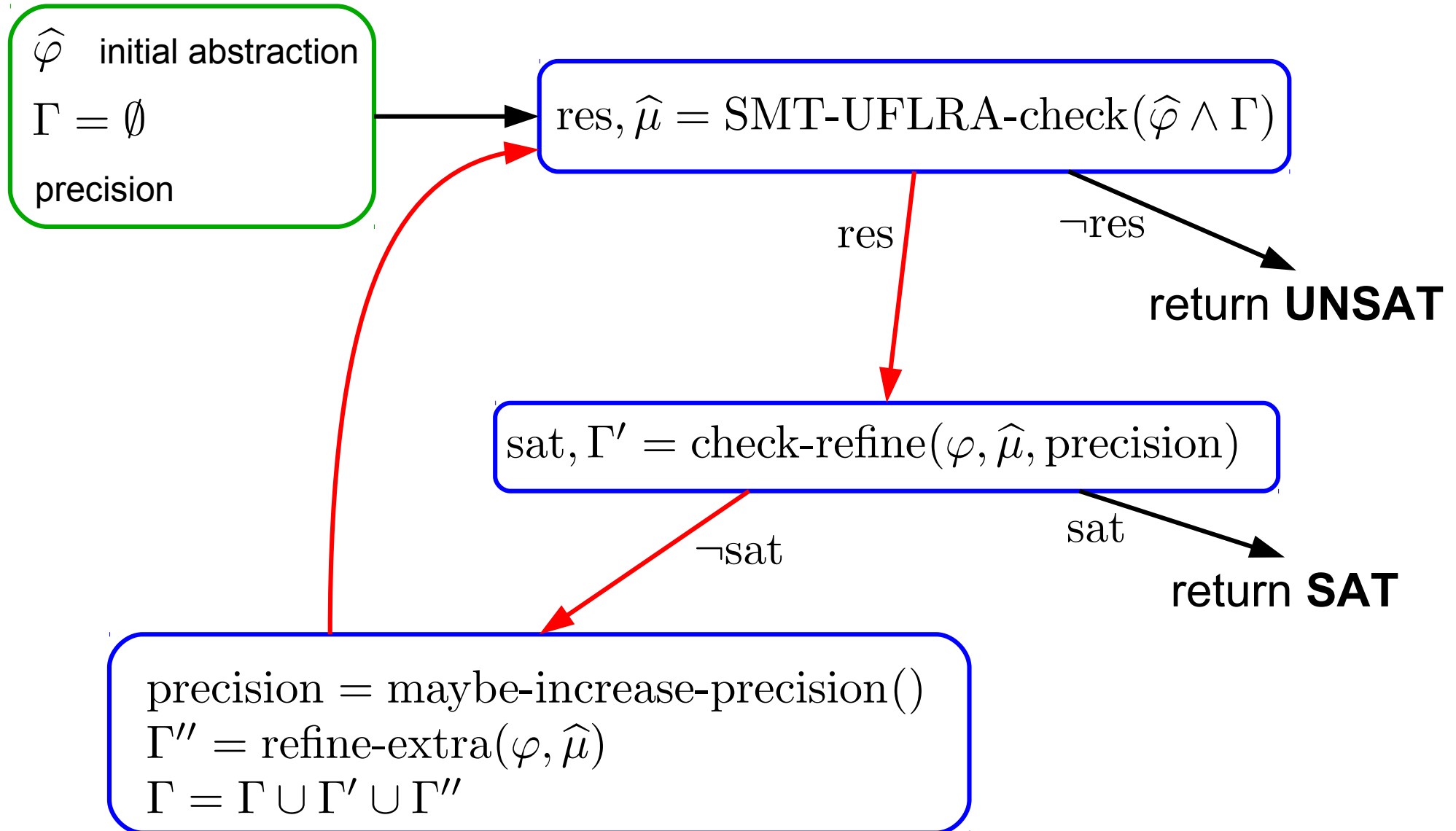
- **Concavity**: sign of the second derivative

- **Taylor theorem**: $f(x) = \underbrace{\sum_{i=0}^n \frac{f^{(i)}(a)}{i!} * (x - a)^i}_{P_{n,f(a)}} + R_{n+1,f(a)}(x)$

$$R_{n+1,f(a)}(x) \leq \underbrace{\max_{c \in [\min(a,x), \max(a,x)]} (|f^{(n+1)}(c)|)}_{\overline{R_{n+1,f(a)}}^u(x)} * \frac{|(x - a)^{n+1}|}{(n + 1)!}$$

$$P_{n,f(a)}(x) - \overline{R_{n+1,f(a)}}^u(x) \leq f(x) \leq P_{n,f(a)}(x) + \overline{R_{n+1,f(a)}}^u(x)$$

Main Algorithm



Initial Abstraction

- Replace every occurrence of a transcendental function $tf(x)$ with a corresponding uninterpreted function $uf(x)$
- Add some basic lemmas about the behaviour of the function
 - E.g. for exponential

$$\exp(x) > 0$$

$$(x = 0 \leftrightarrow \exp(x) = 1) \wedge (x < 0 \leftrightarrow \exp(x) < 1) \wedge (x > 0 \leftrightarrow \exp(x) > 1)$$

$$x = 0 \vee \exp(x) > x + 1$$

Spuriousness Check

- Check that the model is consistent wrt $tf(x)$
 - Intuitively, check $\hat{\mu}[uf(x)] = tf(\hat{\mu}[x])$
- **Problem:** $tf(\hat{\mu}[x])$ is typically irrational
 - Can't check precisely
- **Solution:** check whether $\hat{\mu}[uf(x)]$ is **close enough** to $tf(\hat{\mu}[x])$
 - use Taylor's theorem to compute polynomial bounds

$$P_l(\hat{\mu}[x]) \leq tf(\hat{\mu}[x]) \leq P_u(\hat{\mu}[x])$$

- Depend on the current precision

- Model is definitely spurious if

$$\hat{\mu}[uf(x)] < P_l(\hat{\mu}[x]) \quad \text{or} \quad \hat{\mu}[uf(x)] > P_u(\hat{\mu}[x])$$

Spuriousness Check and Refinement

```
while true:  
  e := 10-precision  
  L := {}
```

Spuriousness Check and Refinement

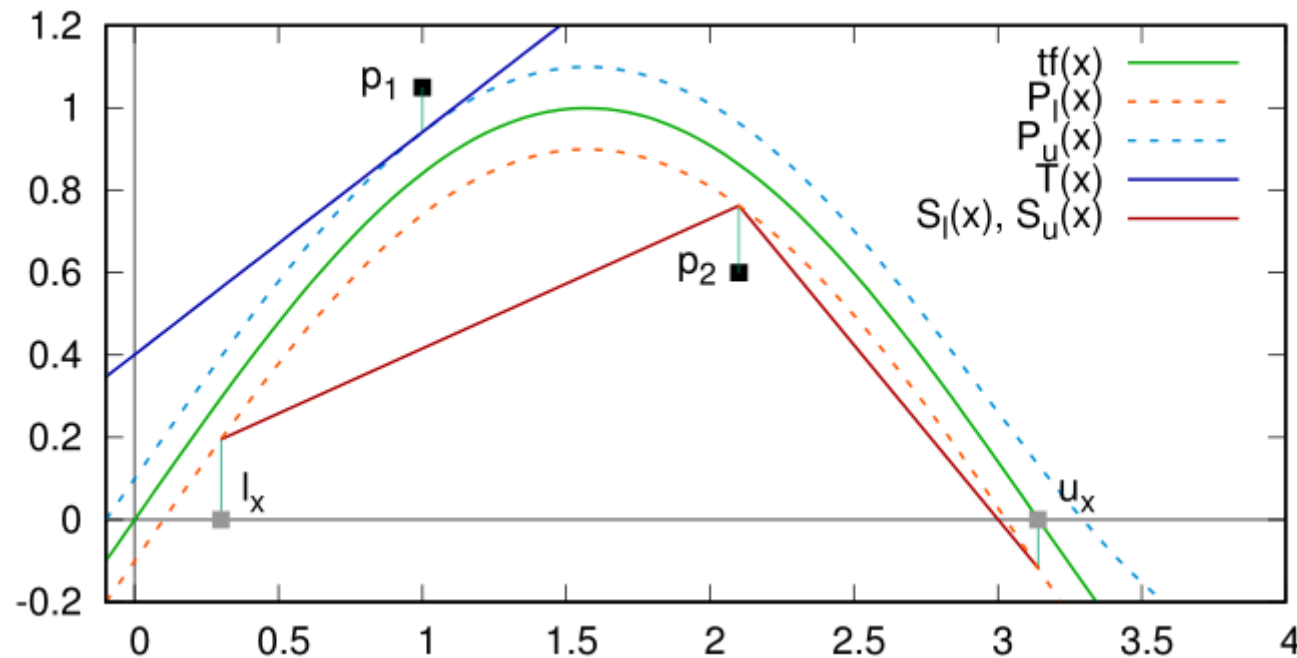
```
while true:
  e := 10-precision
  L := {}
  for all tf(x) in  $\varphi$ :
    c :=  $\hat{\mu}[x]$ 
     $P_l(x), P_u(x) := \text{poly-approx}(tf(x), c, e)$ 
    if  $\hat{\mu}[uf(x)] < P_l(\hat{\mu}[x])$  or  $\hat{\mu}[uf(x)] > P_u(\hat{\mu}[x])$ :
      L := L + get-lemmas-point(tf(x),  $\hat{\mu}$ ,  $P_l(x)$ ,  $P_u(x)$ )
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Spuriousness Check and Refinement

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while true:
    e := 10-precision
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    for all tf(x) in  $\varphi$ :
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        if  $\hat{\mu}[uf(x)] < P_l(\hat{\mu}[x])$  or  $\hat{\mu}[uf(x)] > P_u(\hat{\mu}[x])$ :
            L := L + get-lemmas-point(tf(x),  $\hat{\mu}$ ,  $P_l(x), P_u(x)$ )
    if L is empty:
        if check-sat( $\varphi, \hat{\mu}$ ):
            return true
        else:
            precision += 1
    else:
        return false, L
```

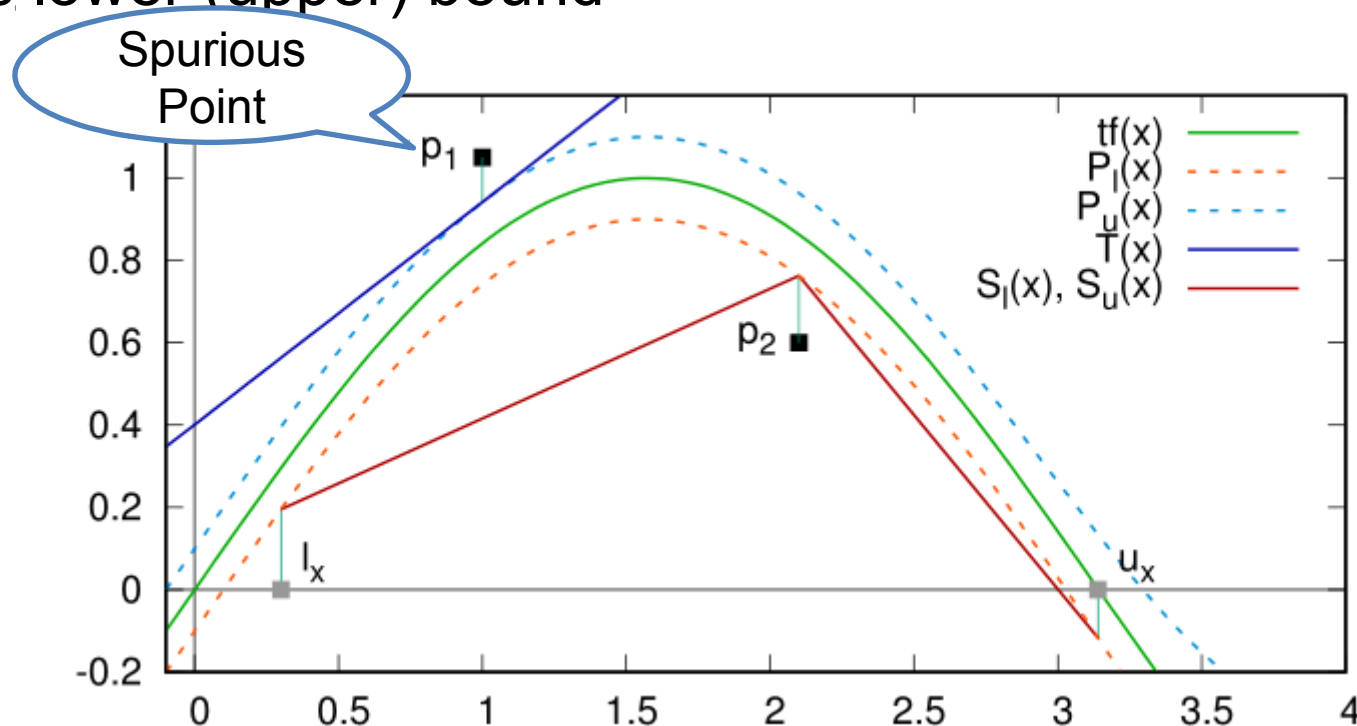
Refinement via Linearization – Basic Idea

- Use upper and lower **polynomials** for linearization
- If the **concavity** of the function is negative (positive):
 - **Tangent Refinement:** Tangent Line to the upper (lower) polynomial gives upper (lower) bound
 - **Secant Refinement:** Secant Line to the lower (upper) polynomial gives lower (upper) bound



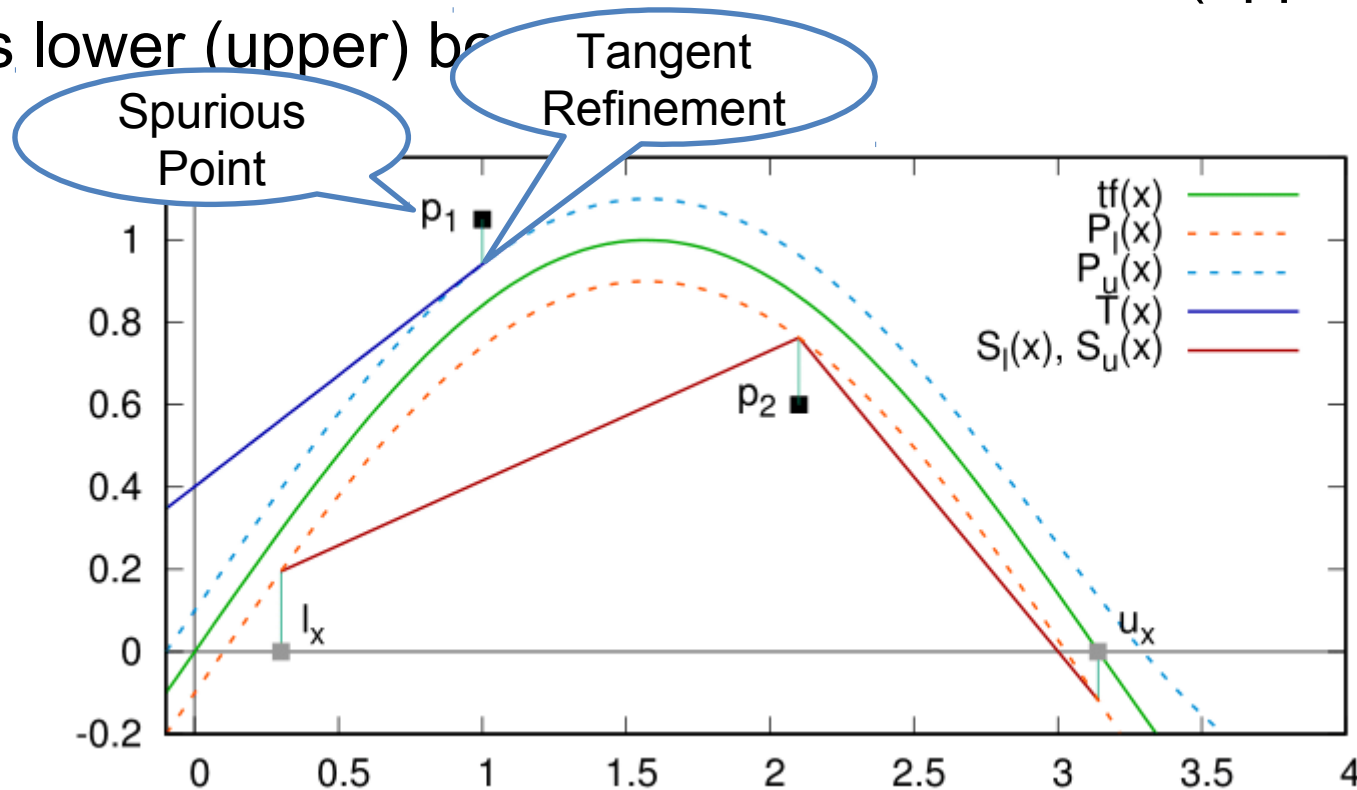
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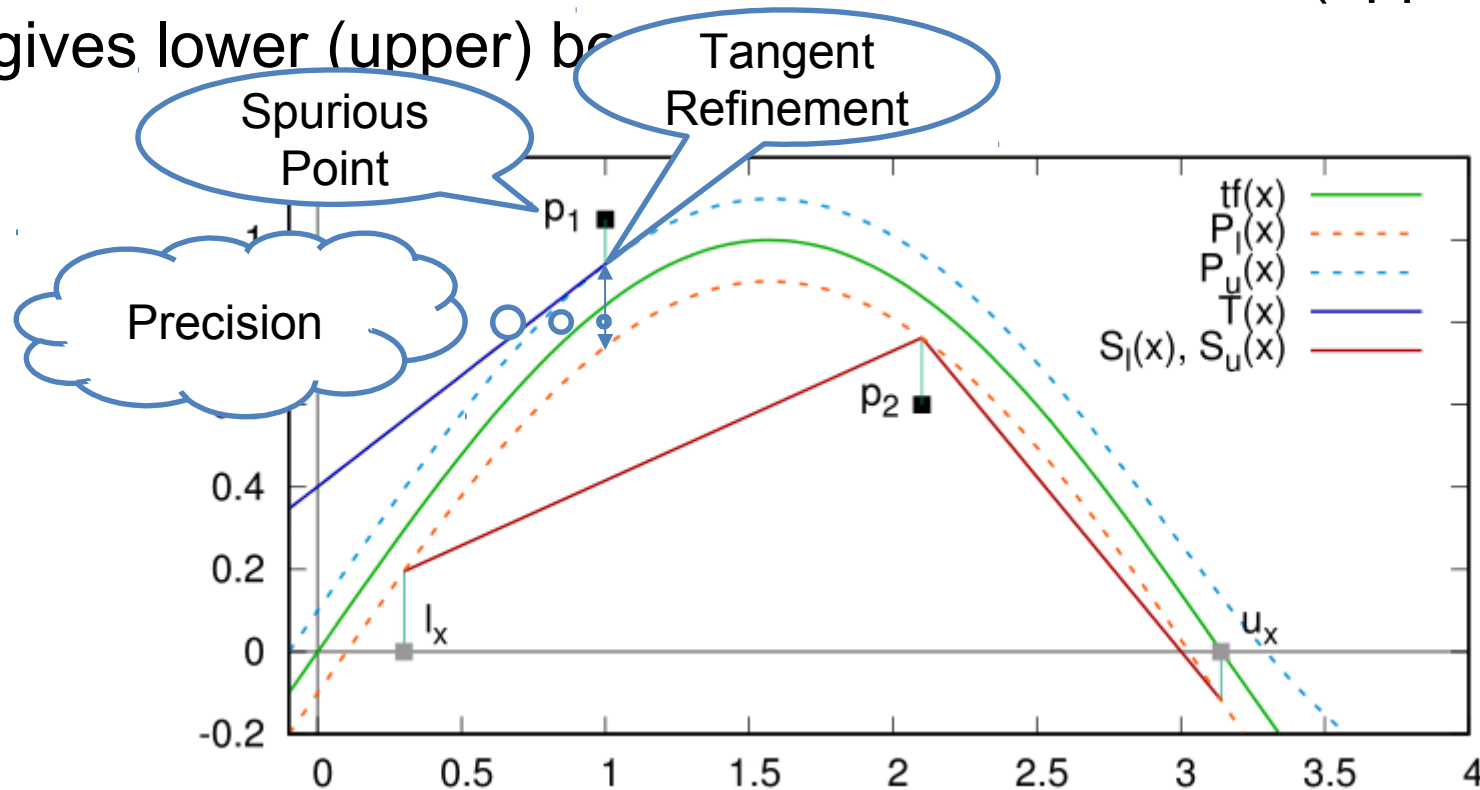
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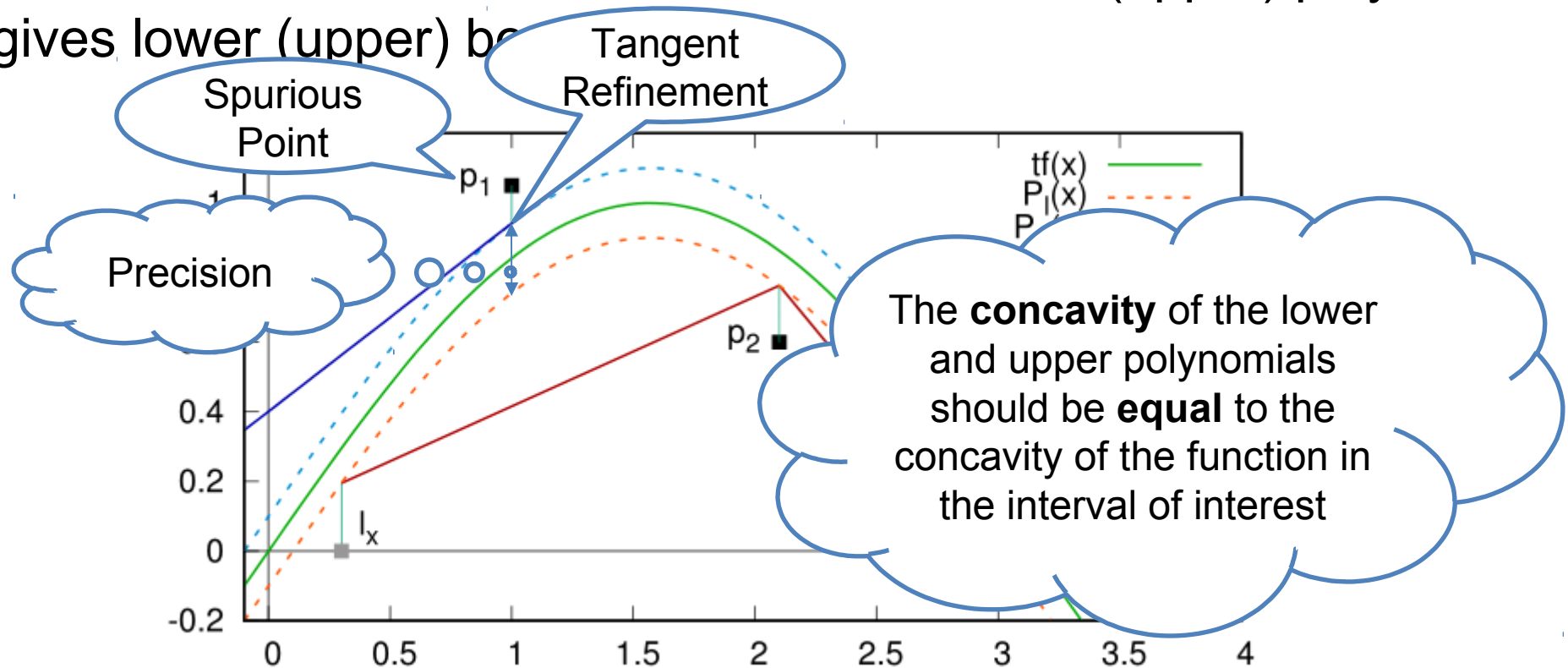
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Refinement: Exponential Function

- Using **Taylor's theorem**:

- **Case $x = 0$:**

- Lower Polynomial: $P_l(0) = 1$

- Upper Polynomial: $P_u(0) = 1$

- **Case $x < 0$:**

- Lower Polynomial: $P_l(x) = \sum_{i=0}^n \frac{x^i}{i!}$ when n is odd

- Upper Polynomial: $P_u(x) = \sum_{i=0}^n \frac{x^i}{i!}$ when n is even

- **Case $x > 0$:**

- Lower Polynomial: $P_l(x) = \sum_{i=0}^n \frac{x^i}{i!}$

- Upper Polynomial: $P_u(x) = \sum_{i=0}^n \frac{x^i}{i!} * \left(1 - \frac{x^{n+1}}{(n+1)!}\right)^{-1}$

Sin Function

- We introduce a **symbolic** $\hat{\pi}$ variable with initial rational bounds

$$\frac{333}{106} < \hat{\pi} < \frac{355}{113}$$

- Reasoning is **split depending on two periods:**

- **Base Period:** $-\hat{\pi}$ to $\hat{\pi}$
- **Extended Period:** when not in the base period

- For each $\sin(x)$, new application $\sin(y_x)$

- y_x fresh (called a **base variable**)
- The domain of y_x is in the **base period**
- $\sin(x)$ and $\sin(y_x)$ are **equal in the base period**

$$(-\hat{\pi} \leq y_x \leq \hat{\pi}) \wedge ((-\hat{\pi} \leq x \leq \pi) \rightarrow y_x = x) \wedge f \sin(x) = f \sin(y_x)$$

Sin Function

- Tangent and secant refinement **only with base variables**
- Concavity check in the base period is easy
 - **Case $x > 0$** : concavity is negative
 - **Case $x < 0$** : concavity is positive
- Using Taylor's theorem and current precision

- **Lower Polynomial:**

$$P_l(y_x) = \sum_{k=0}^n \frac{(-1)^k * y_x^{2k+1}}{(2k+1)!} - \frac{y_x^{2(n+1)}}{(2(n+1))!}$$

- **Upper Polynomial:**

$$P_u(y_x) = \sum_{k=0}^n \frac{(-1)^k * y_x^{2k+1}}{(2k+1)!} + \frac{y_x^{2(n+1)}}{(2(n+1))!}$$

Sin Function – Extended periods

- **Shift** $\hat{\mu}[x]$ to the base period, and compare with $\hat{\mu}[y_x]$
 - Shift calculation $s = (\hat{\mu}[x] + \hat{\mu}[\hat{\pi}]) / (2 \cdot \hat{\mu}[\hat{\pi}])$
- If the values differ, we perform **shift refinement**
 - Relate the extended period with the base period (after appropriate shift)

$$(\hat{\pi} * (2s - 1) \leq x \leq \hat{\pi} * (2s + 1)) \rightarrow (y_x = x - 2s * \hat{\pi})$$

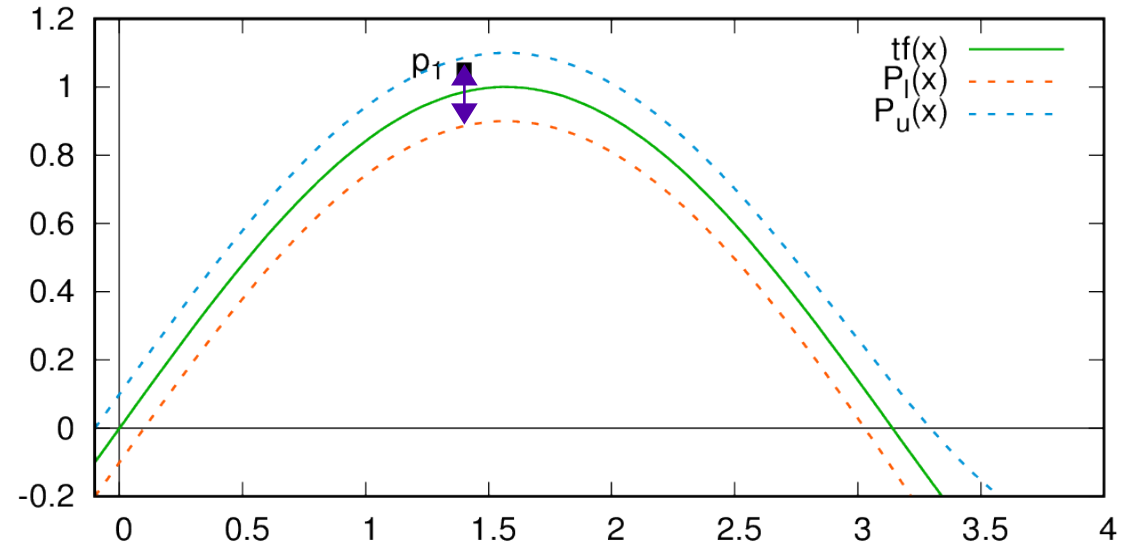
- Note that the shift is **symbolic** in $\hat{\pi}$
 - Ensure soundness

Check for SAT

- We know $\hat{\mu} \models \hat{\varphi}$

$$P_l(\hat{\mu}[x_i]) \leq \hat{\mu}[uf(x_i)] \leq P_u(\hat{\mu}[x_i])$$

so, $\hat{\mu}$ is a candidate solution



- Sufficient condition for sat: validity of

$$\forall uf \in \hat{\varphi}. (\bigwedge_i P_l(\hat{\mu}[x_i]) \leq uf(\hat{\mu}[x]) \leq P_u(\hat{\mu}[x])) \rightarrow \hat{\varphi}\{X \mapsto \hat{\mu}[X]\}$$

Replace $uf(\hat{\mu}[x_i])$'s with fresh vars y_i and check validity of the first-order formula

$$\forall Y. ((\bigwedge_i P_l(\hat{\mu}[x_i]) \leq y_i \leq P_u(\hat{\mu}[x])) \rightarrow \hat{\varphi}\{X \mapsto \hat{\mu}[X]\}) \{Y \mapsto uf(\hat{\mu}[X])\})$$

Implementation and Experiments

- Prototype Implementation in MathSAT + PySMT
- 887 BMC Benchmarks (also scaled)
 - Hand-crafted benchmarks
 - Discretized Hybrid System benchmarks
 - HyComp benchmarks
 - iSAT benchmarks
 - HyST benchmarks
 - HARE benchmarks
- 681 MetiTarski Benchmarks (also scaled)
- 944 dReal Benchmarks
- Tools
 - **MathSAT**
 - MetiTarski
 - iSAT3
 - dReal

Results



Benchmarks	Bounded Model Checking (887)			Mathematical (681)		
Result	SAT	UNSAT	MaybeSAT	SAT	UNSAT	MaybeSAT
METITARSKI	N.A.	N.A.	N.A.	N.A.	530	N.A.
MATHSAT	72	553	N.A.	0	210	N.A.
MATHSAT-NOUNISAT	44	554	N.A.	0	221	N.A.
iSAT3	N.A.	N.A.	N.A.	N.A.	N.A.	N.A.
DREAL	N.A.	392	281 (67/23)	N.A.	285	316 (0/253)
Benchmarks	Scaled Bounded Model Checking (887)			Scaled Mathematical (681)		
Result	SAT	UNSAT	MaybeSAT	SAT	UNSAT	MaybeSAT
MATHSAT	84	556	N.A.	0	215	N.A.
MATHSAT-NOUNISAT	48	556	N.A.	0	229	N.A.
iSAT3	35	470	87 (32/7)	0	212	137 (0/115)
DREAL	N.A.	403	251 (77/23)	N.A.	302	245 (0/195)

Table 1. Results on the BMC and Metitarski benchmarks.

Results

Benchmarks		DREAL (all) (944)			Benchmarks		DREAL (exp/sin only) (96)		
Status		SAT	UNSAT	MaybeSAT	Status		SAT	UNSAT	MaybeSAT
DREAL (orig.)	N.A.	102		524(3/4)	DREAL (orig.)	N.A.	17		37 (3/3)
MATHSAT	3	68		N.A.	MATHSAT	3	39		N.A.
DREAL	N.A.	44		57(3/4)					

Table 2. Results on the Dreal benchmarks.

Thank You

