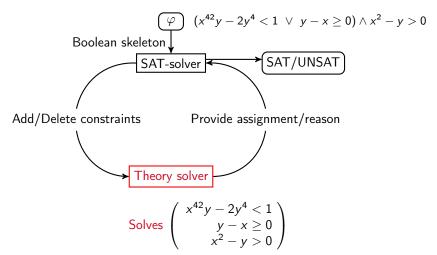
# Satisfiability Checking Subtropical Satisfiability

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WS 19/20

#### Reminder: SMT



The subtropical method is an efficient but incomplete way to check for satisfiability of some of these constraints.

# Subtropical Satisfiability

- Let f be a multivariate polynomial and  $f \sim 0$  with  $\sim \in \{>, \geq, =\}$ .
- The subtropical method quickly finds a solution to  $f \sim 0$  with strictly positive real variables or returns unknown. If a solution is found: The constraint is satisfiable. Else:  $f \sim 0$  might still hold.
- The method is incomplete and thus can be used in conjunction with other techniques.

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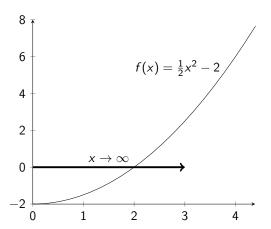
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If the constraint is not an equality, we are done after step 2.

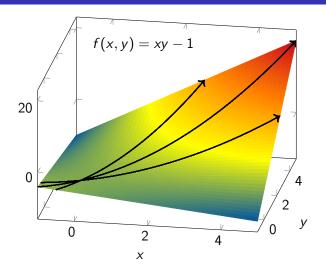
#### Intuition – univariate

We observe for a univariate f(x):

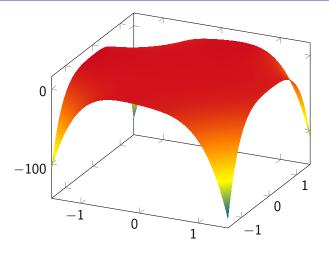
$$\lim_{x\to\infty} f(x) = \left\{ \begin{array}{ll} \infty & \textit{if the leading coefficient is positive} \\ -\infty & \textit{else} \end{array} \right.$$



## Intuition - multivariate 1



#### Intuition – multivariate 2



$$f(x,y) = y + 2xy^3 - 3x^2y^2 - x^3 - 4x^4y^4$$

## Handling multivariate polynomials

- Consider f(x, y) and have (x, y) "growing in some direction".
- For this direction the two variables x and y can be parameterised by a single variable t to construct  $f^*$  (starting from f(1,1)).
- Instead of having f on  $\mathbb{R}^2$ , we consider f restricted to a line as  $f^*$  and have the simple case from before.

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How do we find a "direction" which has a positive leading coefficient?

# The frame of a multivariate polynomial f

#### Definition

```
For f = \sum_{i=1,2,...,n} f_i x_1^{e_i,1} \dots x_d^{e_i,d} \in \mathbb{Z}[x_1,\dots,x_d] with (i) n > 0, (ii) (e_{i,1}, ldots, e_{i,d}) \neq (e_{j,1}\dots, e_{j,d}) for i \neq j and (iii) f_i \neq 0 for i = 1, \dots, n we define: frame(f) = \{(e_{i,1},\dots,e_{i,d}) \mid i \in \{1,\dots,n\}\} frame^+(f) = \{(e_{i,1},\dots,e_{i,d}) \mid i \in \{1,\dots,n\} \land f_i > 0\} frame^-(f) = \{(e_{i,1},\dots,e_{i,d}) \mid i \in \{1,\dots,n\} \land f_i < 0\}
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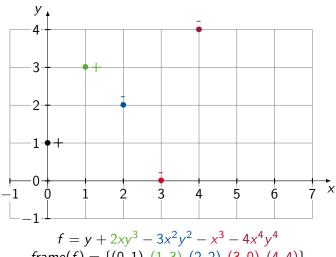
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$$f = y + 2xy^3 - 3x^2y^2 - x^3 - 4x^4y^4$$
  
frame(f) = {(0,1),(1,3),(2,2),(3,0),(4,4)}

And we define based on the signs of the coefficients:

frame<sup>+</sup>(
$$f$$
) = {(0,1), (1,3)}  
frame<sup>-</sup>( $f$ ) = {(2,2), (3,0), (4,4)}

# The frame of a multivariate polynomial f visualized



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# The Newton polytope of a polynomial f

#### Convex hull

Given  $S \subseteq \mathbb{R}^d$ , the *convex hull conv*(S)  $\subseteq \mathbb{R}^d$  is the smallest (inclusion-minimal) convex set containing S.

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#### Convex hull

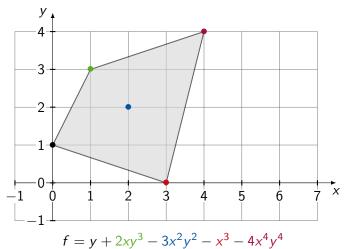
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#### Newton polytope

The Newton polytope of a polynomial  $f \not\equiv 0$  is the convex hull of its frame: Newton(f) = conv(frame(f))

This is indeed a polytope because the convex hull of finite non-empty set of points is bounded.

# The Newton polytope of a polynomial f visualized



The shaded region is the Newton polytope Newton(f) of f.

## Faces of a polytope

Given a polytope  $P \subseteq \mathbb{R}^d$ , the face of P with respect to a vector  $n \in \mathbb{R}^d$  is:  $face(n, P) = \{ p \in P \mid n^T p \ge n^T q \text{ for all } q \in P \}.$ 

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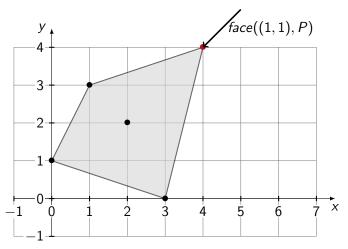
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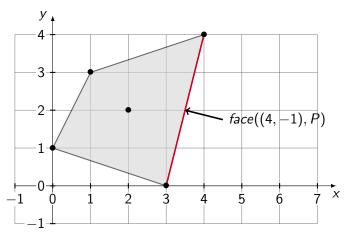
Because frame(f) is finite, it is true that:  $V(Newton(f)) \subseteq frame(f) \subseteq Newton(f)$ ).

# The faces of a Newton polytope visualized



The shaded region is the Newton polytope P of f.  $face((1,1), P) = \{(4,4)\}$  has dimension 0.

# The faces of a Newton polytope visualized



The shaded region is the Newton polytope P of f.  $face((4,-1),P)=\{(3,0)+t(1,4)|0\leq t\leq 1\}$  has dimension 1.

## Hyperplanes separating vertices of the polytope

### Hyperplanes

A *hyperplane* is a subspace whose dimension is one less than that of its surrounding space. Such a *hyperplane* H can be described by the following equation:

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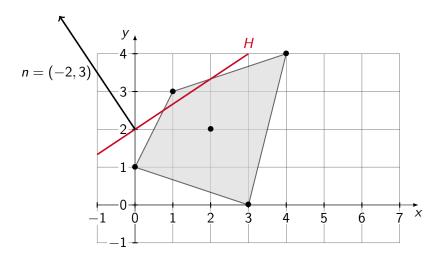
$$H: n^T x + c = 0$$

#### Lemma 1

Let f be a polynomial,  $p \in frame(f)$  and  $n \in \mathbb{R}^d$ . Then the following are equivalent:

- **11**  $p \in V(Newton(f))$  with respect to n.
- 2 There exists  $c \in \mathbb{R}$  such that the hyperplane  $H: n^T x + c = 0$  strictly separates p from  $frame(f) \setminus \{p\}$ . The normal vector n is directed from H towards p.

# Hyperplanes separating vertices of the polytope visualized



The shaded region is the Newton polytope P of f.

 $H: (-2,3)^T x - 6 = 0$  strictly separates (1,3) from the frame(f).

## Vertices as dominating monomials

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#### Lemma 2

Let f be a polynomial and  $p \in frame(f)$  be a vertex of Newton(f) with respect to  $n \in \mathbb{R}^d$ . Then there exists  $a_0 \in \mathbb{R}^+$  such that for all  $a \in \mathbb{R}^+$  with  $a \ge a_0$  the following holds:

- $2 sign(f(a^{n_1},\ldots,a^{n_d})) = sign(f_p)$

where  $f_p$  is the corresponding coefficient to p and  $f_p a^{n^T p} = f_p (a^{n_1})^{p_1} (a^{n_2})^{p_2} ... (a^{n_d})^{p_d}$ .

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To find a point with all positive coordinates where f > 0 we now just need to find a  $p \in frame^+(f)$  and check if it is also a vertex of Newton(f).

## Recap

Let f a multivariate polynomial. If

- $\blacksquare$  *H* is a hyperplane with normal vector n,
- $p \in frame^+(f)$  and
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How do we find n and H?

## Example

Let f again be a multivariate polynomial with:

$$f = y + 2xy^3 - 3x^2y^2 - x^3 - 4x^4y^4$$

- As we have seen the point (1,3) is a vertex of Newton(f) with respect to the normal vector (-2,3).
- Lemma 2 guarantees that  $f(a^{-2}, a^3) > 0$  for sufficiently large positive values of a.
- For example: a = 2,  $f(2^{-2}, 2^3) = \frac{51193}{256}$ .
- Generally, a suitable a can be found by starting with a=2 and doubling a until the constraint is satisfied.

#### The linear problem

- Problem: Given a polynomial f, does a point with all positive coordinates exist with f > 0?
- By Lemma 1 the problem can be reduced to finding a hyperplane  $H: n^T x + c = 0$  separating a  $p \in frame^+(f)$  from  $frame(f) \setminus \{p\}$  where  $frame(f) \subset \mathbb{R}^d$  and  $n \in \mathbb{R}^d$  is a vector pointing from H to p.
- This can be expressed as a linear problem with d + 1 real variables n and c:

$$\varphi(p, frame(f), n, c) = n^T p + c > 0 \land \bigwedge_{q \in frame(f) \setminus p} n^T q + c < 0$$

#### The linear problem: example

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- Encoding :

$$\exists n_x, n_y, c: p_x n_x + p_y n_y + c > 0 \land \bigwedge_{q \in frame(f) \setminus p} q_x n_x + q_y n_y + c < 0$$
 for a given  $p \in frame^+(f)$ .

# The last step (for equalities)

- We have f(1,...,1) < 0.
- And we found a such that  $f(a^n) > 0$  for some direction n.
- What do we need to do now?

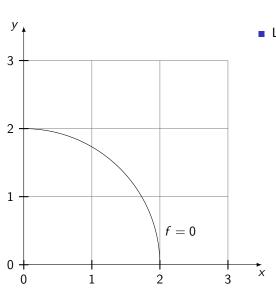
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- Find solution to: f = 0.

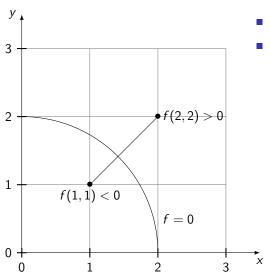
Given: f(1,...,1) < 0 < f(p) for some real coordinates p. Find root of f on the line from (1,...,1) to p.

The Intermediate Value Theorem tells us this root exists.

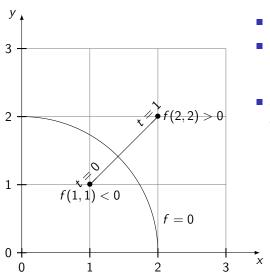
- Construct a new univariate polynomial  $f^*$  from f by parameterising the variables in a new variable t such that we traverse the line from (1,...,1) to p for  $t \in [0,1]$ .
- **2** Find root  $t_0$  of this new polynomial  $f^*$  by common techniques e.g. bisection.
- 3 Construct root of f as point on the line from (1, ..., 1) to p for parameter  $t_0$ .



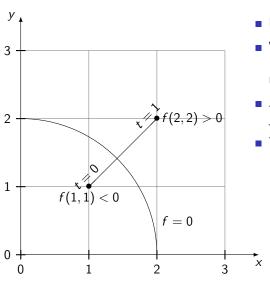
■ Let  $f = x^2 + y^2 - 4$ .



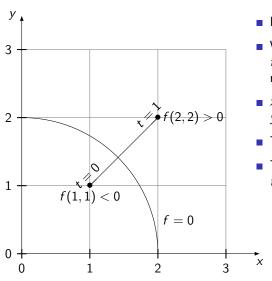
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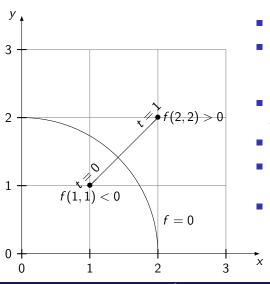


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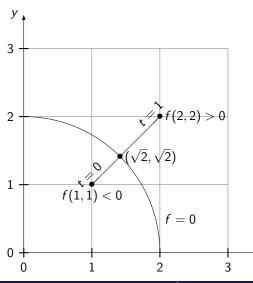
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- Here we have:  $t = \sqrt{2} 1$ .
- $(1 + (\sqrt{2} 1), 1 + (\sqrt{2} 1)) = (\sqrt{2}, \sqrt{2})$  is a root of f.

For constraints of the form f > 0 (or f ≥ 0):
 Solve the linear problem of finding a hyperplane separating a p ∈ frame<sup>+</sup>(f) from the rest of frame(f).
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Not shown here: deal with multiple constraints at once. How? Search for one direction that works for all constraints.