

# Abstractions of Sequences, Functions and Operators

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# Introduction

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- Infer information about **mathematical functions** given by a **recursive definition**, i.e. by an operator/equation.
- Role of **order theory** and **abstract interpretation**.
- Share some order theory facts, Galois connections, ... discovered along the way.
- Present (domain specific?) abstract domains built using **functions as the basic object**.

# Intuition

## Idea to be explored

Functions  $X \rightarrow Y$  may be simpler objects than arbitrary sets of points  $\mathcal{P}(X \times Y)$  (relations).

Of course, not in general, but...

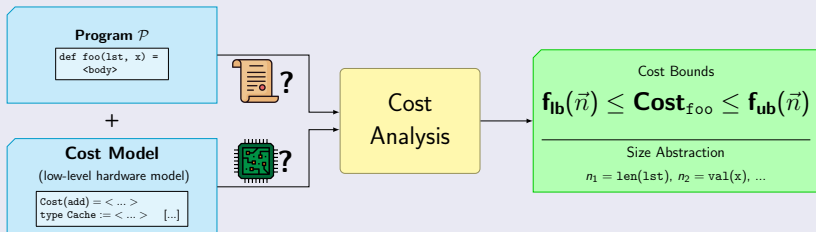
- Can be the case for families of functions we care to approximate
- Exploit local regularity
- Relate definition with that of simpler functions
- ...

## Analysis viewpoint

We are interested in **numerical functions**,  
**constructed**/defined recursively **by an operator**  
 $\Phi \in ((X \rightarrow L) \rightarrow (X \rightarrow L))$ , **i.e. by an equation.**

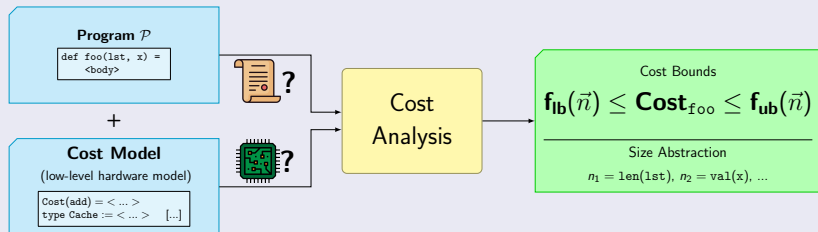
# Origin and Motivation: Cost Analysis

## Cost Analysis: Bounds on Resource Consumption



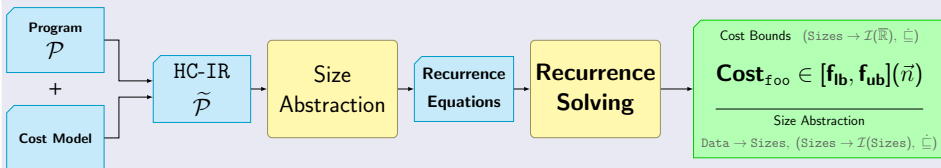
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- Pipeline implemented in  PP (and other analysers).

## Recurrence-based cost analysis



# Intuition

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- Solution of a **differential equation**, ...

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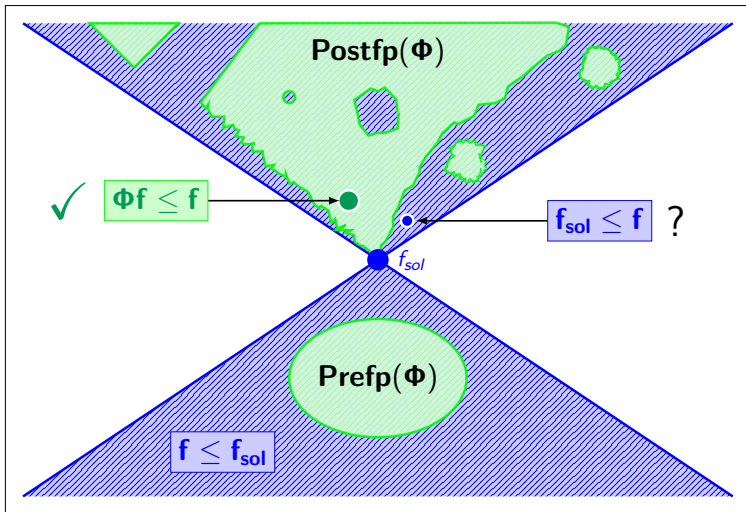
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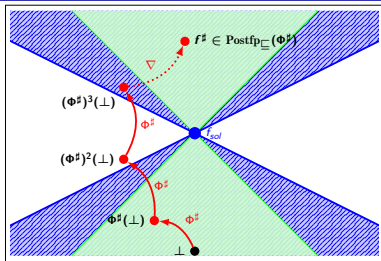
→ Can we build abstract lattices of functions, and  
abstract these operators/equations ( $\Phi \rightsquigarrow \Phi^\#$ ) to  
**produce bounds**  $f_{sol} \sqsubseteq \hat{f}$  by **abstract Kleene iteration**?



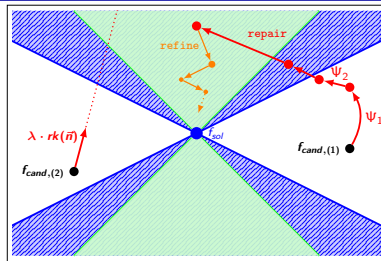
# Order theory framework of equations [SAS24]

Equations  $\leftrightarrow$  OperatorsSolutions  $\leftrightarrow$  FixpointsBounds  $\leftarrow$  Pre/Postfixp

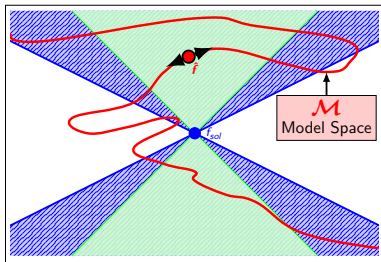
# Equation solving as pre/postfixpoint search [SAS24]



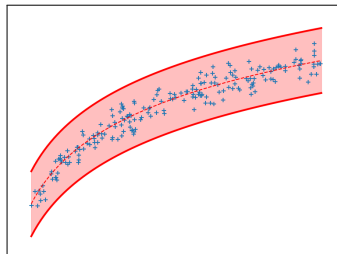
Abstract Interpretation



Geometry-based expression Repair

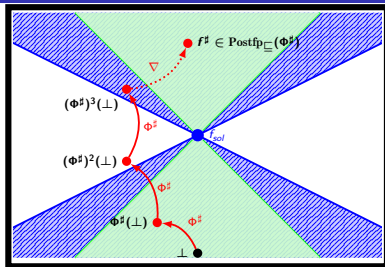


Search on subvarieties: **Templates**,  $\forall$ -elim

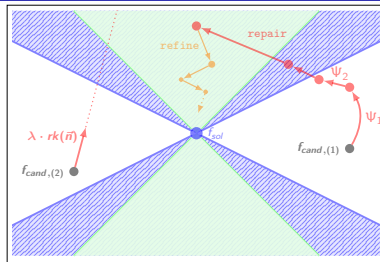


Constrained **Optimisation**,  
with *provability constraints*

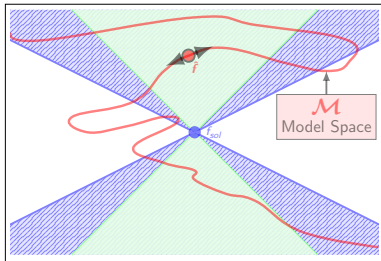
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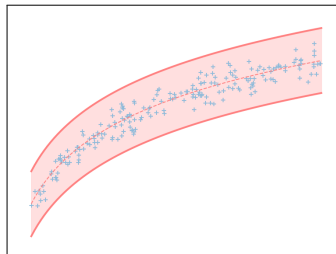
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# (Abstract) Lattices of Functions and Galois Connections between them

# Lattices of Functions

## Pointwise lattice structure

For  $X$  a set and  $(L, \leq, \vee, \wedge)$  a complete lattice, the set of functions  $X \rightarrow L$  has a very natural lattice structure  $(X \rightarrow L, \dot{\leq}, \dot{\vee}, \dot{\wedge})$ , with

$$f \dot{\leq} g \iff \forall x \in X, f(x) \leq g(x),$$

$$f \dot{\vee} g \triangleq (x \mapsto f(x) \vee g(x)), \quad \dot{\perp} \triangleq (x \mapsto \perp),$$

$$f \dot{\wedge} g \triangleq (x \mapsto f(x) \wedge g(x)), \quad \dot{\top} \triangleq (x \mapsto \top).$$

## Example (Numerical functions)

In  $(\mathcal{D} \rightarrow \overline{\mathbb{R}}, \dot{\leq})$ , join and meet are pointwise max/min.

$$\dot{\perp} = x \mapsto -\infty.$$

## Example (Set-valued functions)

In  $(\mathcal{D} \rightarrow \mathcal{I}(\mathbb{R}), \dot{\subseteq}_{\mathcal{I}})$ , join/meet are pointwise union/intersection.

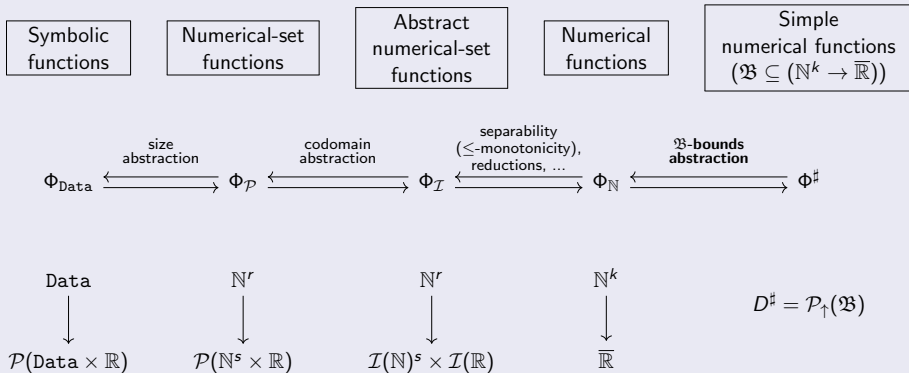
We are interested in objects which **can be viewed as functions**, constructed/defined as **fixpoints** of some **operator**  $\Phi \in ((X \rightarrow L) \rightarrow (X \rightarrow L))$ , which is **monotone** for such an order on functions (to apply Knaster-Tarski).

- **Function defined recursively** in a declarative language,
- **Input/output of a basic block** in an imperative language,
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→ Can we abstract these lattices  $(D \xrightarrow[\alpha]{\gamma} D^\#)$  and these operators  $(\Phi \rightsquigarrow \Phi^\#)$  to produce bounds  $f_{sol} \dot{\subseteq} \hat{f} \in \text{Postfp}(\Phi)$  by **abstract Kleene iteration**?

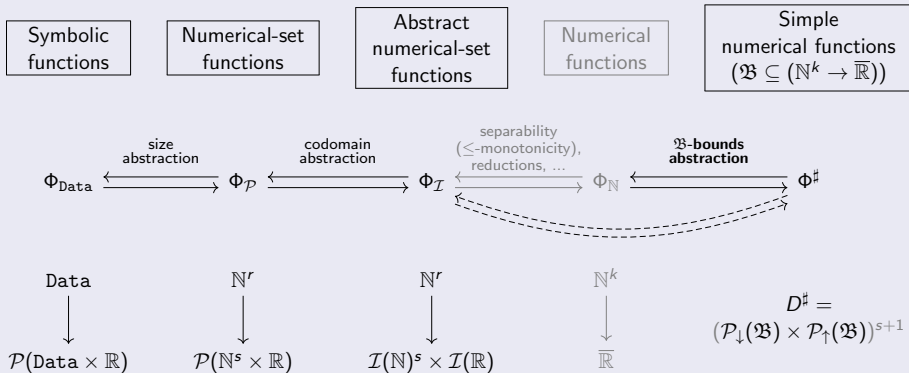
# Overview: a landscape of abstractions

## From symbolic functions to manageable numerical functions



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## From symbolic functions to manageable numerical functions



**Simplified picture:** intervals for numerically non-monotone systems, lb+ub, systems  $\Phi \in \text{End}(\prod_{\tau} (\tau_{in} \rightarrow \tau_{out}))$  vs single equation  $\Phi \in \text{End}(\tau_{in} \rightarrow \tau_{out})$ , ...

# Constructing Galois connections in function space

Many ways to build connections on functions from connections on values.

## Proposition (Codomain abstraction)

Let  $(D, \leq) \xleftrightarrow[\alpha]{\gamma} (D^\sharp, \sqsubseteq)$  and  $X$  be a set. This lifts to  $(X \rightarrow D, \dot{\leq}) \xleftrightarrow[\dot{\alpha}]{\dot{\gamma}} (X \rightarrow D^\sharp, \dot{\sqsubseteq})$ ,  
with  $\dot{\alpha}(f) = \alpha \circ f$  and  $\dot{\gamma}(f^\sharp) = \gamma \circ f^\sharp$ .



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## Definition (Endomorphisms are monotone endofunctions)

Let  $(L, \sqsubseteq)$  be a partial order.  $\text{End}_{\sqsubseteq}(L) := \{f : L \rightarrow L \mid \forall x \sqsubseteq y, f(x) \sqsubseteq f(y)\}$ .

## Proposition (End-lifting)

Let  $(D, \leq) \xleftrightarrow[\alpha]{\gamma} (D^\#, \sqsubseteq)$ . This lifts to a Galois connection

$$\begin{aligned} (\text{End}_{\leq}(D), \dot{\leq}) &\xleftrightarrow[\dot{\alpha}]{\dot{\gamma}} (\text{End}_{\sqsubseteq}(D^\#), \dot{\sqsubseteq}) \\ f &\mapsto \alpha \circ f \circ \gamma, \\ \gamma \circ f^\# \circ \alpha &\longleftarrow f^\#. \end{aligned}$$

$$\begin{array}{ccc} D & \xrightarrow{f} & D \\ \gamma \uparrow & & \downarrow \alpha \\ D^\# & \xrightarrow[\text{---}]{f^\# = \dot{\alpha}(f)} & D^\# \end{array}$$

## Corollary

This can be iterated to **operators** in  $\text{End}^2 \approx ((\cdot \rightarrow \cdot) \rightarrow (\cdot \rightarrow \cdot))$ , and beyond.

# From mappings in value space to Galois connections in function space

## Theorem (Domain abstraction)

Let  $m : X \rightarrow A$  be an **arbitrary mapping**,  
and  $(L, \sqsubseteq)$  be a complete lattice.

Then, there is a Galois connection

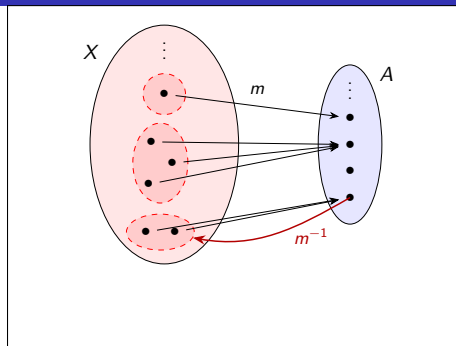
$$(X \rightarrow L, \dot{\sqsubseteq}) \xleftrightarrow[\alpha]{\gamma} (A \rightarrow L, \dot{\sqsubseteq})$$

$$f \mapsto \left( a \mapsto \bigsqcup_{x \in m^{-1}(a)} f(x) \right)$$

$$\left( x \mapsto f^\#(m(x)) \right) \longleftarrow f^\#,$$

which is an insertion for  $m : X \rightarrow A$ .

“ $\alpha : f \mapsto \bigsqcup f \circ m^{-1}$ ,  $\gamma : f^\# \mapsto f^\# \circ m$ ”



**Remark:** Codomain abstraction is classical.

$$(X \rightarrow D, \dot{\sqsubseteq}) \xleftrightarrow{\quad} (X \rightarrow D^\#, \dot{\sqsubseteq})$$

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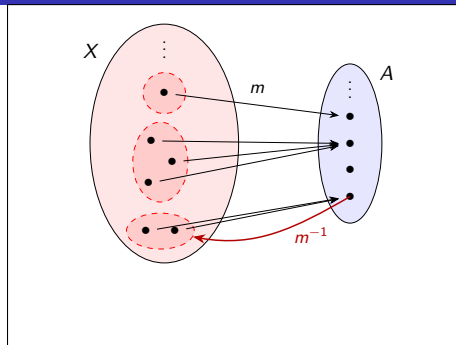
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## Example (Size abstraction)

$m : \text{Data} \rightarrow \text{Sizes}$

e.g. list-length  $\sqsubseteq$  tree-nbnodes  $\sqsubseteq$  int-value...

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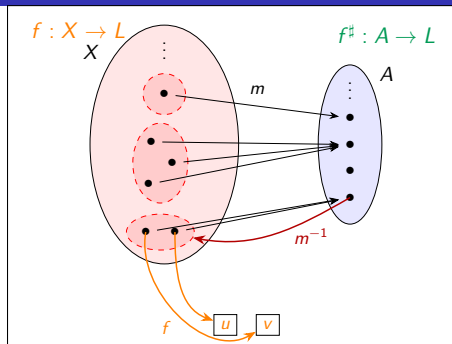
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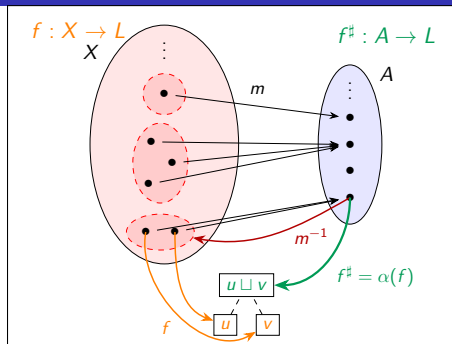
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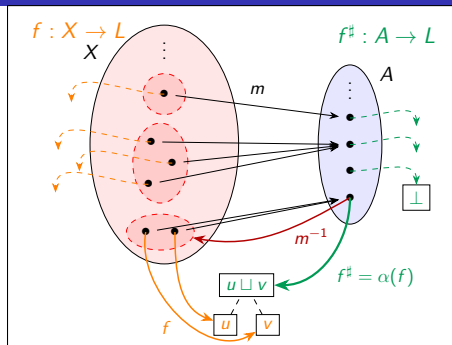
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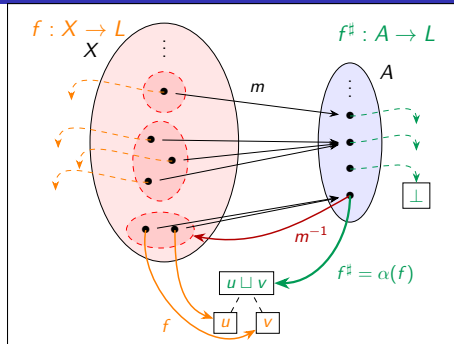
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## Corollary (Powerset lifting)

$$(\mathcal{P}(X), \subseteq) \overset{\gamma}{\underset{\alpha}{\rightleftarrows}} (\mathcal{P}(A), \subseteq)$$

## Corollary (For non-deterministic functions)

$$(X \rightarrow \mathcal{P}(X), \overset{\gamma}{\underset{\alpha}{\rightleftarrows}} (A \rightarrow \mathcal{P}(A), \overset{\gamma}{\underset{\alpha}{\rightleftarrows}} \\ ("f \mapsto m^{-1} \circ f \circ m \text{ and } f^\# \mapsto m \circ f^\# \circ m^{-1}").$$

## Example (Size abstraction)

$m : \text{Data} \rightarrow \text{Sizes}$

e.g. list-length  $\sqcup$  tree-nbnodes  $\sqcup$  int-value...

- Can also combine such *size metrics*.
- Question of inference for a given program.
- $\approx$ Dual approach: Sized-Types  $\rightarrow \mathcal{P}(\text{Data})$ .

**Remark:** Codomain abstraction is classical.

$$(X \rightarrow D, \leq) \rightleftarrows (X \rightarrow D^\#, \overset{\gamma}{\underset{\alpha}{\rightleftarrows}})$$

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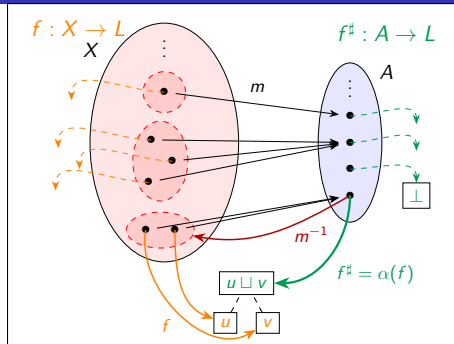
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## Example (Dimensionality reduction)

$$m : \mathbb{N}^3 \rightarrow \mathbb{N}^2 \\ (x, y, z) \mapsto (x + y, z)$$

Important questions:

- Inference for a given program  $\Phi$ ,
- Computation of  $\Phi^\#$ , ...

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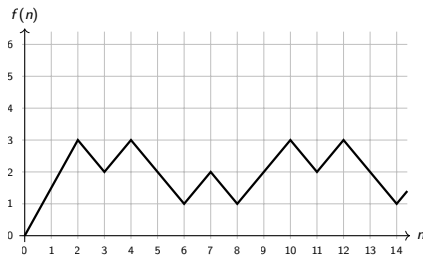
# $\mathfrak{B}$ -bounds — abstracting functions with simpler functions

## A class of abstract domains for functions

For  $\mathfrak{B} \subseteq (\mathcal{D} \rightarrow \overline{\mathbb{R}})$ , let  $D^\# := (\mathcal{P}(\mathfrak{B}), \supseteq)$  be the domain of  $\mathfrak{B}$ -upper bounds.

$(\mathcal{D} \rightarrow \overline{\mathbb{R}}, \dot{\leq}) \xleftrightarrow[\alpha_{\mathfrak{B}}]{\gamma_{\mathfrak{B}}} (\mathcal{P}(\mathfrak{B}), \supseteq)$ . For  $f : \mathcal{D} \rightarrow \overline{\mathbb{R}}$ ,

$$\alpha_{\mathfrak{B}}(f) = F_{ub}^\# \iff \bigwedge_{f_i^\# \in F_{ub}^\#} f \dot{\leq} f_i^\#.$$



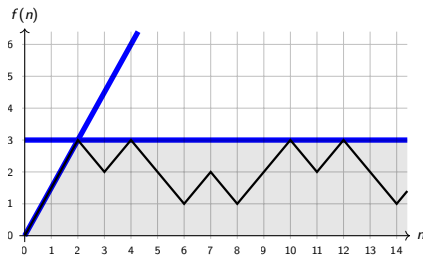
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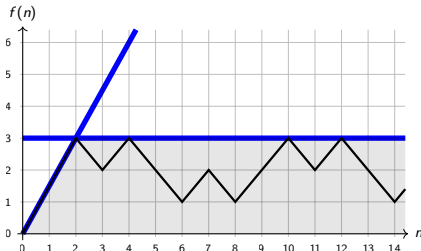
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## Examples

- **Affine bounds.**  $\mathfrak{B} = \{n \mapsto an + b \mid a, b \in \overline{\mathbb{R}}\}$  (one dim),  $n \mapsto \vec{a} \cdot \vec{n} + b$  (multidim).
- **Polynomial bounds (bounded degree).**  $n \mapsto \sum_{k \leq d} a_k n^k$  (monomial basis),  
 $n \mapsto \sum a_k \binom{n}{k}$  (binomial basis), multidim versions, ...
- **Poly-exp.**  $n \mapsto \sum a_{b,k} b^n n^k$ , or  $n \mapsto \sum a_{b,k} \binom{n}{b+1}^{k+1}$  (with Stirling numbers of 2nd kind), ...
- **Arithmetico-geometric sequences, Regular expressions on numbers,**  $\sum a b^n n^k \log(en + f) \dots$
- **Extra features:** initial exactness, piecewise behaviour (disjunctive versions), ...

## Remark: functional version of a familiar concept (for relations) – constraint domains

For  $E \in \mathcal{P}(\mathbb{R}^{\text{Vars}})$ ,

$$\alpha_P(E) = \text{Polyhedron} \leftrightarrow (\forall x \in E, \bigwedge_i \sum_j \alpha_{i,j} x_j \leq \beta_i),$$

$$\alpha_B(E) = \text{Box} \leftrightarrow (\forall x \in E, \bigwedge_i a_i \leq x_i \leq b_i), \quad \alpha_Z(E) = \text{Zone} \leftrightarrow (\forall x \in E, \bigwedge_{i,j} x_i - x_j \leq c_{i,j}), \dots$$

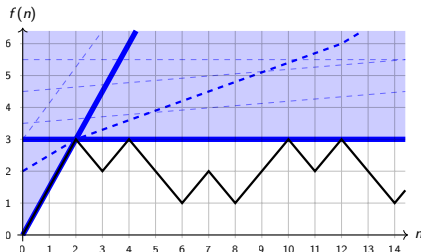
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## Galois connection – $\mathfrak{B}$ -ubs

$$(\mathcal{D} \rightarrow \overline{\mathbb{R}}, \dot{\leq}) \xleftrightarrow[\alpha_{\mathfrak{B}}]{\gamma_{\mathfrak{B}}} (\mathcal{P}(\mathfrak{B}), \supseteq)$$

$$f \mapsto \{f_{ub}^\# \in \mathfrak{B} \mid f \dot{\leq} f_{ub}^\#\}$$

$$\left( \vec{n} \mapsto \min_{f_{ub}^\# \in F_{ub}^\#} f_{ub}^\#(\vec{n}) \right) \longleftarrow F_{ub}^\#$$

In practice, we replace  $(\mathcal{P}(\mathfrak{B}), \supseteq, \cap, \cup)$  by a more computable representation  $(\mathcal{P}_{\uparrow, \text{fin}}(\mathfrak{B}), \sqsubseteq^\#, \sqcup^\#, \cap^\#)$ , with  $A \sqsubseteq^\# B$  a sound approximation of  $\uparrow A \supseteq \uparrow B$ .

## Remarks

- We can restrict to up-closed sets  $\mathcal{P}_\uparrow(\mathfrak{B}) := \{F \subseteq \mathfrak{B} \mid \forall f \in F, \forall g \in \mathfrak{B}, f \dot{\leq} g \Rightarrow g \in F\}$ .
- Up-closure ( $\uparrow$ ) =  $\alpha_{\mathfrak{B}} \gamma_{\mathfrak{B}} : \mathcal{P}(\mathfrak{B}) \rightarrow \mathcal{P}_\uparrow(\mathfrak{B})$ .
- Search for a finite number of **generators**  
 $\uparrow \{f_1^\#, \dots, f_k^\#\} = \alpha_{\mathfrak{B}}(f)$ ,  
 (or at least an overapproximation of  $\alpha_{\mathfrak{B}}(f)$ ).
- Add normalise and widening operators to keep the representation bounded.

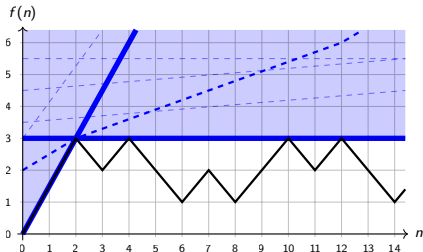
# $\mathfrak{B}$ -bounds — abstracting functions with simpler functions

## A class of abstract domains for functions

For  $\mathfrak{B} \subseteq (\mathcal{D} \rightarrow \overline{\mathbb{R}})$ , let  $D^\sharp := (\mathcal{P}(\mathfrak{B}), \supseteq)$  be the domain of  $\mathfrak{B}$ -upper bounds.

$(\mathcal{D} \rightarrow \overline{\mathbb{R}}, \leq) \xleftrightarrow[\alpha_{\mathfrak{B}}]{\gamma_{\mathfrak{B}}} (\mathcal{P}(\mathfrak{B}), \supseteq)$ . For  $f : \mathcal{D} \rightarrow \overline{\mathbb{R}}$ ,

$$\alpha_{\mathfrak{B}}(f) = F_{ub}^\sharp \iff \bigwedge_{f_i^\sharp \in F_{ub}^\sharp} f \leq f_i^\sharp.$$



## With intervals – Galois connection with $\mathfrak{B}$ -bounds: “flowpipes”

$$(\mathcal{D} \rightarrow \mathcal{I}(\mathbb{R}), \sqsubseteq_{\mathcal{I}}) \xleftrightarrow[\alpha_{\mathfrak{B}}]{\gamma_{\mathfrak{B}}} (\mathcal{P}(\mathfrak{B}), \supseteq) \times (\mathcal{P}(\mathfrak{B}), \supseteq)$$

$$\left( \vec{n} \mapsto \left[ \max_{f_{lb}^\sharp \in F_{lb}^\sharp} f_{lb}^\sharp(\vec{n}), \min_{f_{ub}^\sharp \in F_{ub}^\sharp} f_{ub}^\sharp(\vec{n}) \right] \right) \longleftarrow (F_{lb}^\sharp, F_{ub}^\sharp)$$

$$(f_{lb}, f_{ub}) \mapsto \left( \begin{array}{l} \{f_{lb}^\sharp \in \mathfrak{B} \mid \forall \vec{n} \in \mathcal{D}, f_{lb}^\sharp(\vec{n}) \leq f_{lb}(\vec{n})\}, \\ \{f_{ub}^\sharp \in \mathfrak{B} \mid \forall \vec{n} \in \mathcal{D}, f_{ub}^\sharp(\vec{n}) \geq f_{ub}(\vec{n})\} \end{array} \right)$$

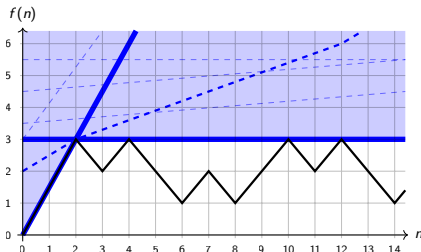
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## Proposition (convexity of constraint set)

**When  $\mathfrak{B}$  is convex,**  
**all  $\alpha(f)$  are convex (in function space).**

## Example

Most of the templates  $\mathfrak{B}$  discussed before are convex, e.g.

$$\begin{aligned} & \lambda \cdot (n \mapsto \sum a_k n^k) + (1 - \lambda) \cdot (n \mapsto \sum b_k n^k) \\ &= (n \mapsto \sum (\lambda a_k + (1 - \lambda) b_k) n^k). \end{aligned}$$

→ This reduces the problem of finding a minimal set of extremal bounds (e.g. for transfer function synthesis) to finding generators of a convex set.

To make the problem tractable in practice, we often approximate the problem  $\Phi(\gamma(f^\#)) \dot{\leq} g^\#$ , and **work in parameter space**, using an order that is an incomplete abstraction of  $\dot{\leq}$ , e.g.

$$((n \mapsto \sum a_k n^k) \sqsubseteq^\# (n \mapsto \sum b_k n^k)) \xleftrightarrow{\Delta} (\forall k, a_k \leq b_k), \quad (\text{despite } n^2 - 1 \not\geq_{\mathbb{N}} 0).$$

# $\mathfrak{B}$ -bounds: discussion

We have discussed Galois connections, and mentioned that the domain can support non-linear bounds.

Let's make this concrete, with fully-fledged abstract domains:

- add a **language**, (we will present an operator language)
- design **transfer functions**,
- showcase **abstract iteration**.

→ Let's build some  $\mathfrak{B}$ -bounds domains!

---

Other applications of  $\mathfrak{B}$ -bound domains, beyond abstract iteration:

- Simplify analysis output for users:  $f \leq g_{\text{precise}} \rightsquigarrow \bigwedge_i f \leq g_{i,\text{readable}}$ ,
- Help decide inequality of functions:  $f \dot{\leq}^{(?)} g$ ,
- ...

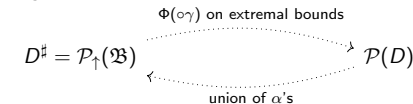
# Abstract Domains of Sequences: A simple way towards non-linearity



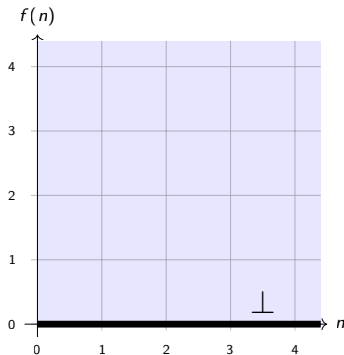
# Abstract iteration – what we want to get

Equation  $\Phi(f) = \text{ite}(n > 0, f(f(n-1)) + 1, 0)$ ,  $D = \mathbb{N} \rightarrow \mathbb{N}_\infty$ ,  $\mathfrak{B} = \text{Affines}$ .

→ We deal with extremal bounds separately.



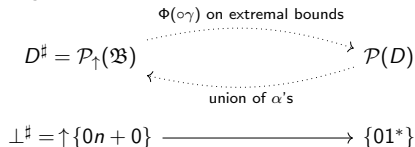
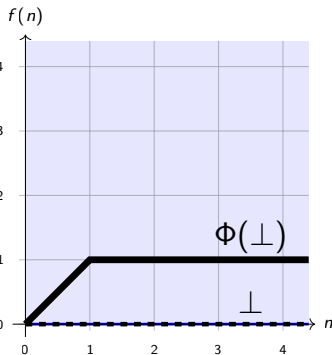
$$\perp^\# = \uparrow\{0n + 0\}$$



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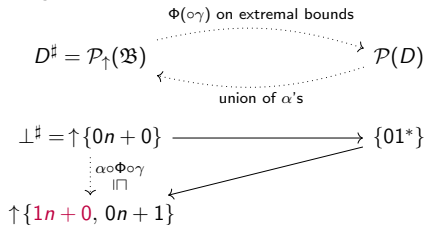
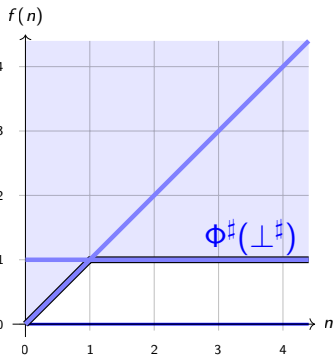
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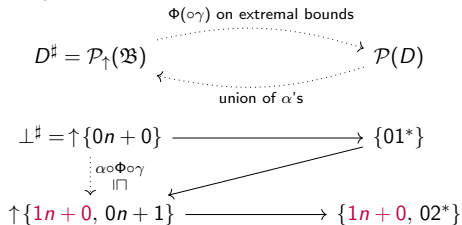
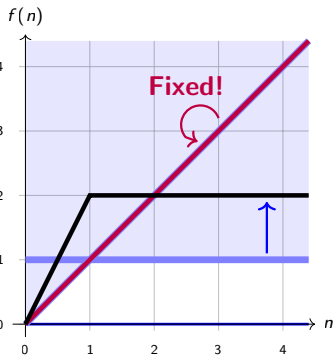
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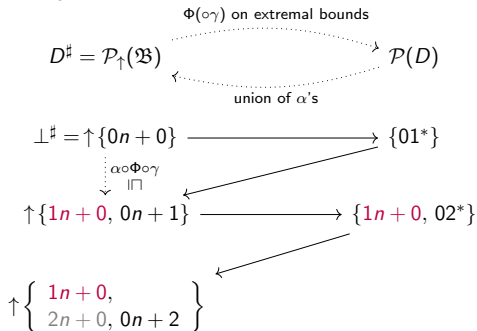
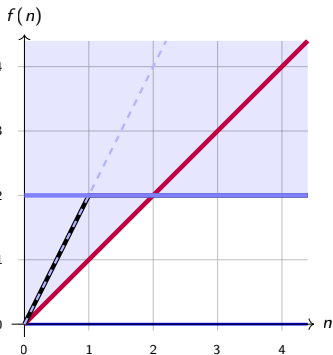
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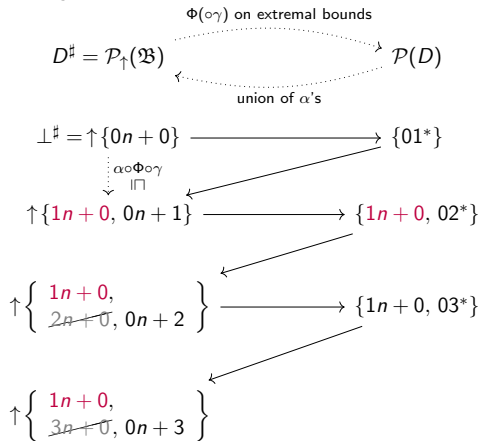
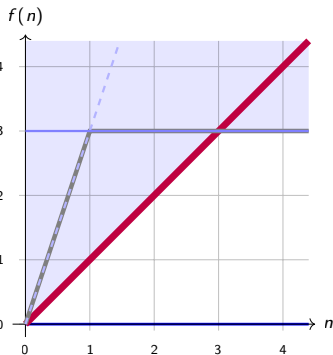
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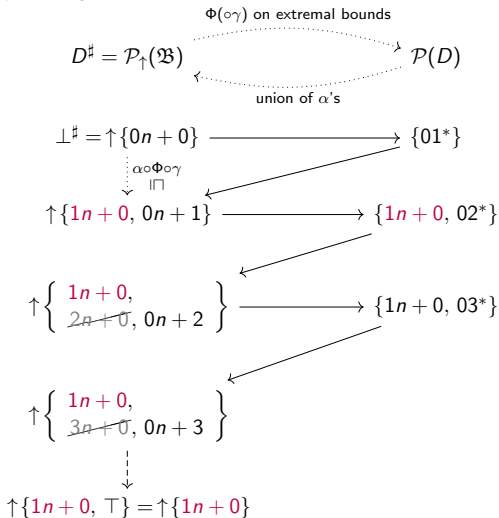
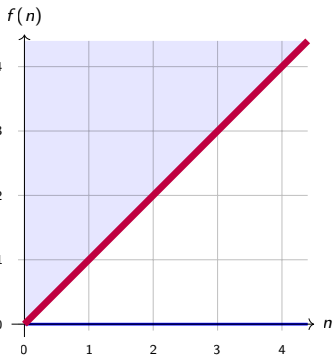
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→ We deal with extremal bounds separately.



# Sequences

We start with an easy case: one dim., simple recursive structure (push/pop).

## Syntax

$\langle \text{expr} \rangle ::=$   
 $\quad | \text{Cst } \langle \text{num} \rangle \mid 'n' \mid 'f'$   
 $\quad | \langle \text{expr} \rangle \diamond \langle \text{expr} \rangle$   
 $\quad | \text{Pop } \langle \text{expr} \rangle \mid \text{Push } \langle \text{num} \rangle \langle \text{expr} \rangle$   
 $\quad | \dots$   
 $\langle \text{num} \rangle ::= c \in \overline{\mathbb{R}} \quad \diamond \in \{+, -, \times\}$

Seq: A simple operator language

## Example

$f(0) = 4, f(n) = 1/2 \cdot f(n-1) + 3$   
 $\longleftrightarrow \Phi : f \mapsto n \mapsto \text{ite}(n = 0, 4, 1/2 \cdot f(n-1) + 3)$   
 $\longleftrightarrow \text{Push } 4 \ (\text{Cst}(1/2) \times ('f' + \text{Cst}(3)))$

## Semantics

$$\llbracket \cdot \rrbracket : \text{Seq} \rightarrow \text{End}(\mathbb{N} \rightarrow \overline{\mathbb{R}})$$

$$\llbracket \text{Cst } c \rrbracket(f) = n \mapsto c$$

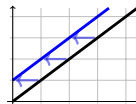
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$$\llbracket e_1 \diamond e_2 \rrbracket(f) = \llbracket e_1 \rrbracket(f) \diamond \llbracket e_2 \rrbracket(f)$$

$$\llbracket \text{Pop} \rrbracket(f) = n \mapsto f(n+1)$$

$$\llbracket \text{Push } c \rrbracket(f) = n \mapsto \text{ite}(n = 0, c, f(n-1))$$



Pop



Push c



# Sequences

We start with an easy case: one dim., simple recursive structure (push/pop).

## Syntax

```

<expr> ::=
  | Cst <num> | 'n' | 'f'
  | <expr> ◇ <expr>
  | Pop <expr> | Push <num> <expr>
  | Shift <...> | Scale <...> | ...
<num> ::= c ∈ ℝ    ◇ ∈ {+, −, ×} ∪ {o}
  
```

Seq: A simple operator language

## Example

```

f(0) = 4, f(n) = 1/2 · f(n − 1) + 3
↔ Φ : f ↦ n ↦ ite(n = 0, 4, 1/2 · f(n − 1) + 3)
↔ Push 4 (Cst(1/2) × ('f' + Cst(3)))
  
```

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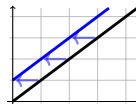
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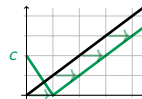
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Pop



Push c

# Warm-up: affine bounds

$$\mathfrak{B} = \{n \mapsto an + b \mid a \in \mathbb{R}_+ \cup \{+\infty\}, b \in \mathbb{R}_+\} \cup \{\top_{\mathfrak{B}}\}$$

For simpler presentation, here only ub, and positive coefficients (need lb for subtraction and multiplication by negatives).

Here, **coordinate order is complete** ( $a_1n + b_1 \leq a_2n + b_2 \iff (a_1, b_1) \sqsubseteq_{\mathfrak{B}} (a_2, b_2)$ )

$$\begin{aligned} (a_1, b_1) \sqsubseteq_{\mathfrak{B}} (a_2, b_2) &\iff a_1 \leq a_2 \wedge b_1 \leq b_2, & \perp_{\mathfrak{B}} &= (0, 0) & \top_{\mathfrak{B}} &\approx (-, +\infty) \\ \bigsqcup \{(a_i, b_i)\} &= (\max_i a_i, \max_i b_i), & \bigsqcap \{(a_i, b_i)\} &= (\min_i a_i, \min_i b_i), \end{aligned}$$

For simplicity, define transfer functions  $\mathfrak{B} \rightarrow \mathcal{P}(\mathfrak{B})$ , then extend to  $\mathcal{P}(\mathfrak{B}) \rightarrow \mathcal{P}(\mathfrak{B})$ .

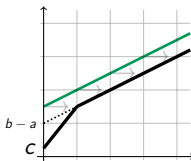
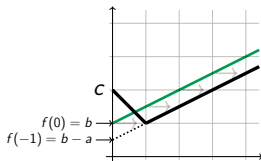
- Constants, addition, multiplication easily represented, e.g.

$$(a_1n + b_1) + (a_2n + b_2) = (a_1 + a_2)n + (b_1 + b_2)$$

$$(a_1n + b_1) \times (a_2n + b_2) = \underbrace{a_1a_2}_{a_1a_2n^2}n + (b_1b_2)$$

- $\llbracket \text{Pop} \rrbracket^{\sharp}((a, b)) = \{(a, b + a)\}$

- $\llbracket \text{Push } c \rrbracket^{\sharp}((a, b)) = \begin{cases} \{(a, c)\} & \text{if } c \geq b - a \\ \{(a, b - a), (b - c, c)\} & \text{if } c < b - a \end{cases}$



## Convex problem for $\llbracket \text{Push } c \rrbracket^{\sharp}$

$$(n \mapsto un + v) \geq \llbracket \text{Push } c \rrbracket(n \mapsto an + b)$$

$$\iff$$

$$\begin{cases} v \geq c \\ \wedge \llbracket \text{Pop} \rrbracket(n \mapsto un + v) \geq (n \mapsto an + b) \end{cases}$$

$$\iff$$

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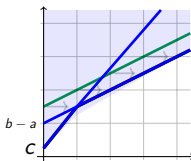
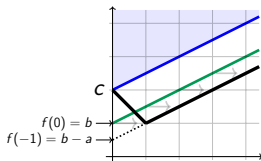
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# Polynomial bounds

$$\mathfrak{B} = \left\{ n \mapsto \sum_{k \leq d} a_k n^k \mid a_k \in \mathbb{R}_+ \right\} \cup \{ a_0(+\infty)^* \mid a_0 \in \mathbb{R}_+ \} \cup \{ \top_{\mathfrak{B}} \}$$

- Similar idea here: **polynomials are closed by simple arithmetic operations** (+, ×, o, ...), and **representations of these operations are easy to compute**.
- The coordinate-wise order becomes incomplete (e.g.  $1 \cdot n^2 + 2 \cdot n \not\leq 2 \cdot n^2 + 1 \cdot n$ ), but this is okay and we can still decide  $\leq$  if necessary.

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- Pop and Push are a bit more complex, but this is manageable.

$$\text{Pop} \quad \sum_{k \leq d} a_k (n+1)^k = \sum_{k \leq d} \underbrace{\left( \sum_{j \leq d} a_k \binom{j}{k} \right)}_{p_k} n^k$$

Push x ??

Convex problem for  $\llbracket \text{Push } x \rrbracket^\sharp$

$$(n \mapsto \sum p_k n^k) \geq \llbracket \text{Push } x \rrbracket (n \mapsto \sum a_k n^k)$$

$$\iff$$

$$\left\{ \begin{array}{l} p_0 \geq x \\ \wedge \quad \forall k \in \llbracket 0, d \rrbracket, \sum_{j \leq d} p_k \binom{j}{k} \geq a_k \end{array} \right.$$

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→ There is a nice trick, coming from works on type-based amortised cost analysis [HoffmannPhd11]: **Use the binomial basis instead of the monomial basis**, which has much simpler behaviour under shifts.

# Polynomial bounds – Binomial basis

$$\mathfrak{B} = \left\{ n \mapsto \sum_{k \leq d} a_k \binom{n}{k} \mid a_k \in \mathbb{R}_+ \right\} \cup \{ a_0(+\infty)^* \mid a_0 \in \mathbb{R}_+ \} \cup \{ \top_{\mathfrak{B}} \}$$

- Instead of the basis  $1, n, n^2, n^3, \dots$  of polynomials, use the basis  $\binom{n}{0}, \binom{n}{1}, \binom{n}{2}, \binom{n}{3}, \dots$   
Examples:  $\binom{n}{0} = 1$ ,  $\binom{n}{1} = n$ ,  $\binom{n}{2} = \frac{1}{2}n^2 - \frac{1}{2}n$ ,  $\binom{n}{3} = \frac{1}{6}n^3 - \frac{1}{2}n^2 + \frac{1}{3}n$ ,
- We can convert between the two representation easily if needed  
(e.g. multiplication, composition, rescaling, etc., can be easier in monomial basis).
- Key identity: 
$$\binom{n+1}{k} = \binom{n}{k} + \binom{n}{k-1}.$$
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Push  $x$  ...

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$$(n \mapsto \sum p_k n^k) \geq \llbracket \text{Push } x \rrbracket (n \mapsto \sum a_k n^k)$$

$\Longleftarrow$

$$\begin{cases} p_0 \geq x \\ \wedge \\ p_d \geq a_d \\ \wedge \quad \forall k < d, p_k + p_{k+1} \geq a_k \end{cases}$$

# Polynomial bounds – Binomial basis

$$\mathfrak{B} = \left\{ n \mapsto \sum_{k \leq d} a_k \binom{n}{k} \mid a_k \in \mathbb{R}_+ \right\} \cup \{ a_0(+\infty)^* \mid a_0 \in \mathbb{R}_+ \} \cup \{ \top_{\mathfrak{B}} \}$$

- Instead of the basis  $1, n, n^2, n^3, \dots$  of polynomials, use the basis  $\binom{n}{0}, \binom{n}{1}, \binom{n}{2}, \binom{n}{3}, \dots$   
Examples:  $\binom{n}{0} = 1$ ,  $\binom{n}{1} = n$ ,  $\binom{n}{2} = \frac{1}{2}n^2 - \frac{1}{2}n$ ,  $\binom{n}{3} = \frac{1}{6}n^3 - \frac{1}{2}n^2 + \frac{1}{3}n$ ,
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Push  $x$  e.g., for  $d = 2$ ,  $\{(p_2, p_1, p_0)\} =$

$$\begin{cases} \{(a_2, a_1 - a_2, x), (x + a_1 - a_0, a_0 - x, x)\} & \text{if } x > a_0 - a_1 + a_2 \\ \{(a_2, a_1 - a_2, a_0 - a_1 + a_2)\} & \text{if } x = a_0 - a_1 + a_2 \\ \{(a_2, a_1 - a_2, a_0 - a_1 + a_2), (a_2, a_0 - x, x)\} & \text{if } x < a_0 - a_1 + a_2 \end{cases}$$

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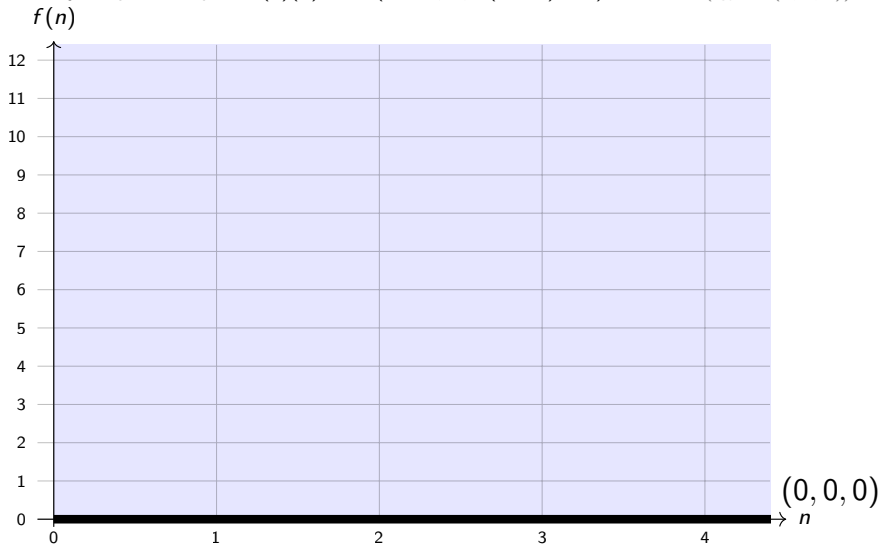
$$\begin{cases} p_0 \geq x \\ \wedge \\ p_d \geq a_d \\ \wedge \quad \forall k < d, p_k + p_{k+1} \geq a_k \end{cases}$$

→ Extends well to higher dimensions, with multivariate polynomials.

# Polynomial bounds – Example

Very simple example:  $\Phi(f)(n) = \text{ite}(n = 0, 0, f(n-1) + n)$ .

$(f_{\text{sol}} = (1, 1, 0))$

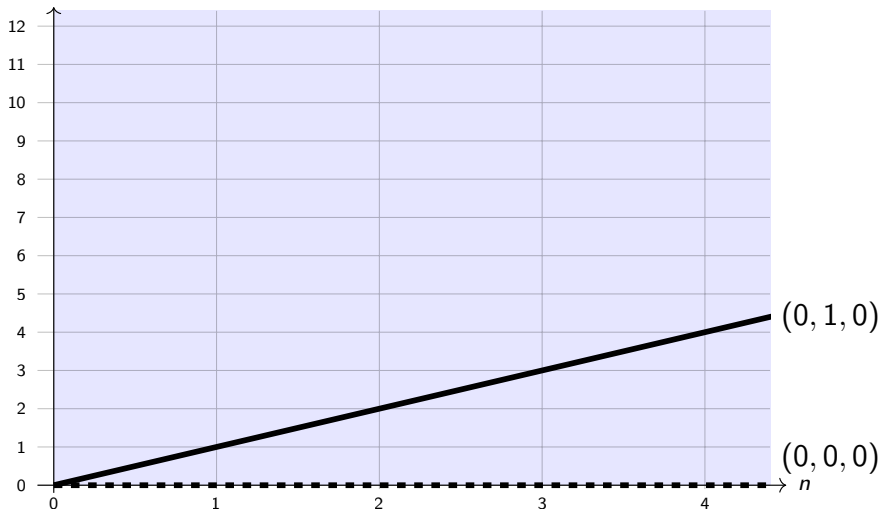


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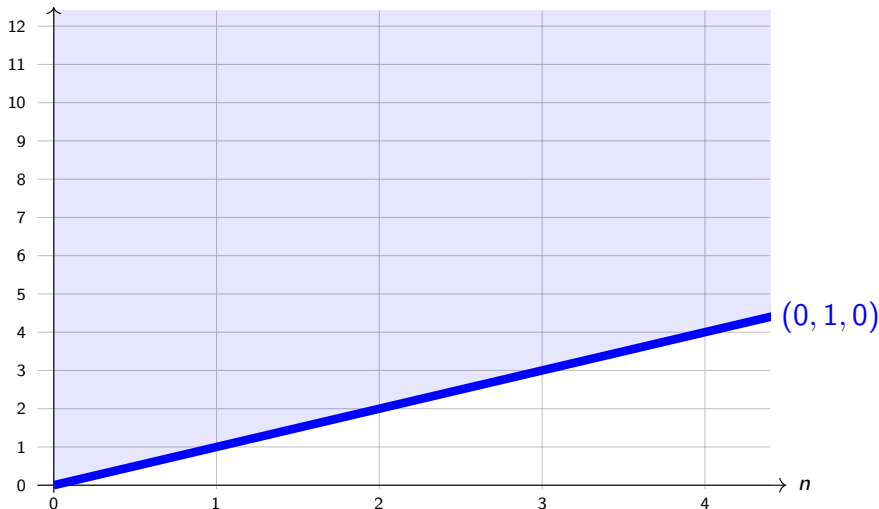


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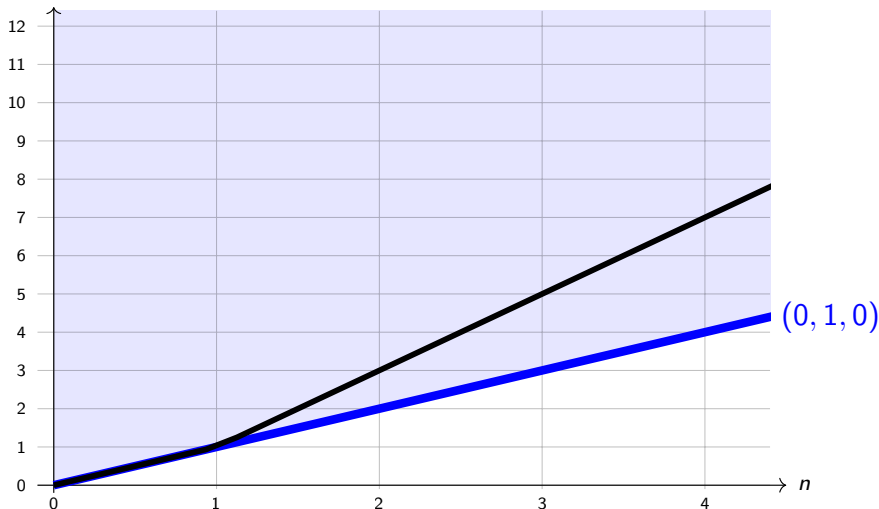


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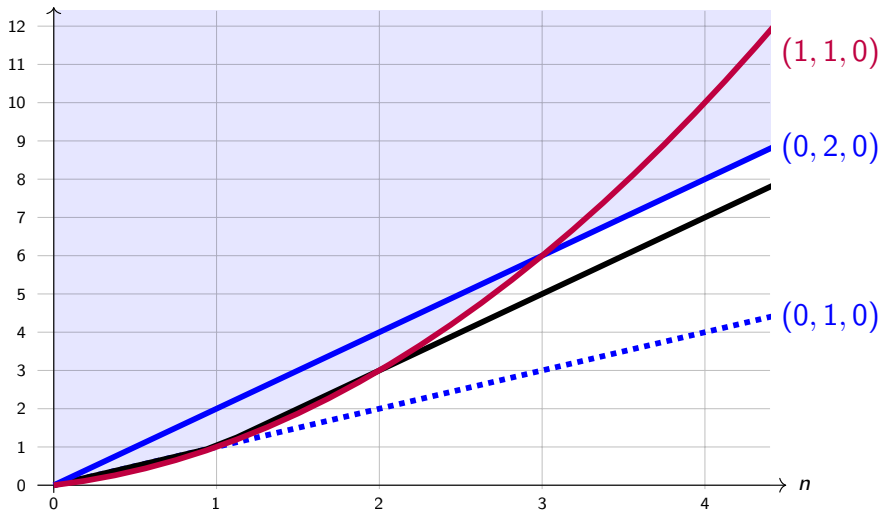


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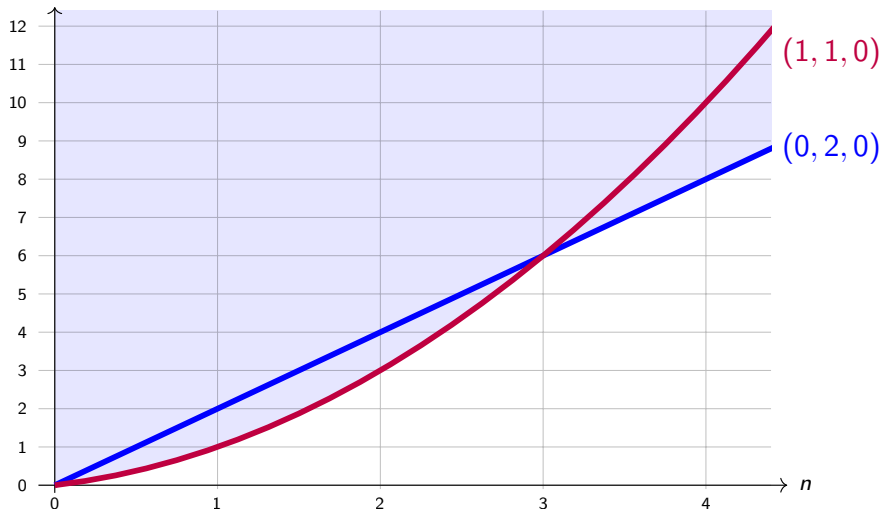


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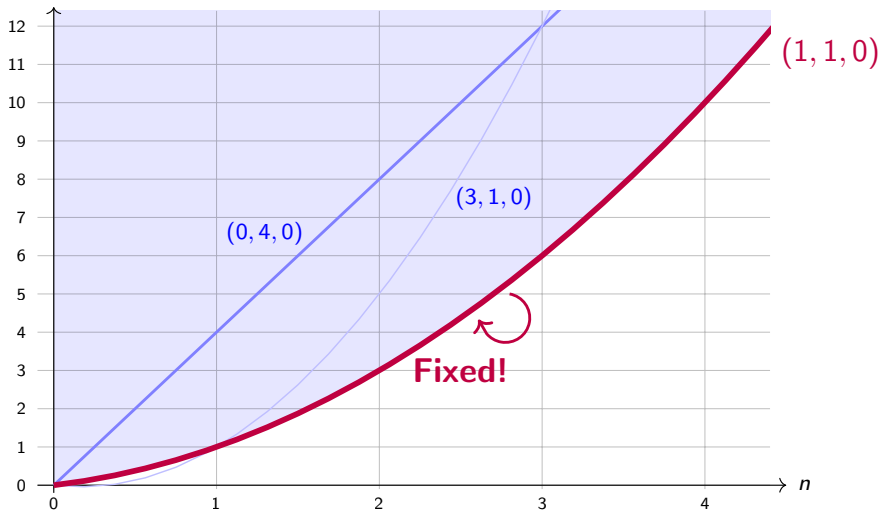


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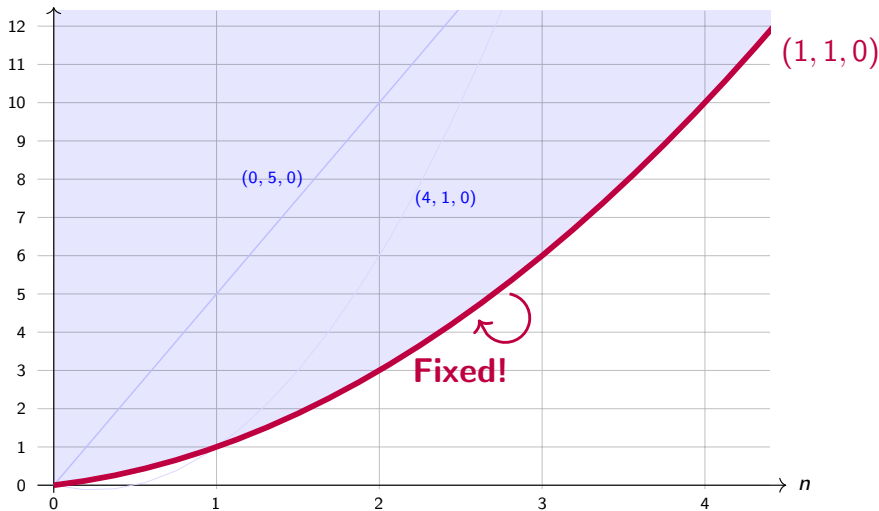


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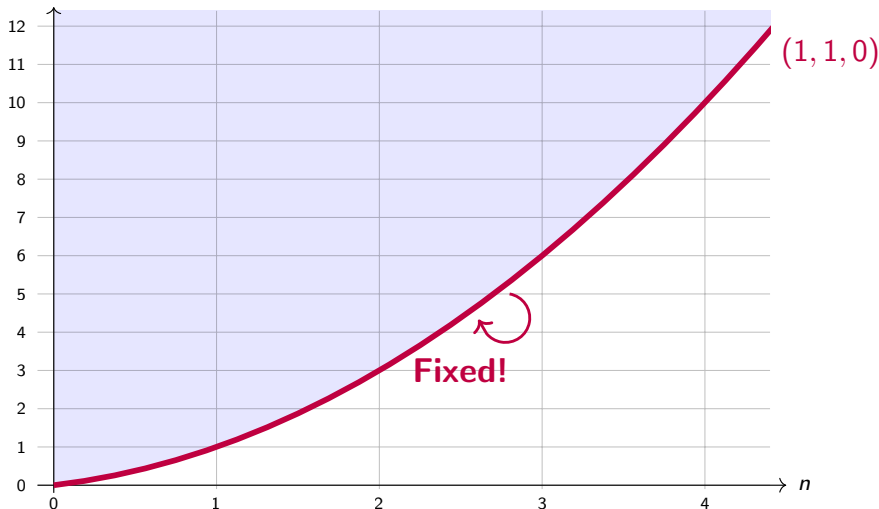


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# Beyond polynomials – exponential polynomials

Adding exponentials: the domain of **arithmetico-geometric sequences** [Feret05].

$$\beta : \mathbb{R}_+^5 \rightarrow \mathfrak{B} \subseteq (\mathbb{N} \rightarrow \mathbb{R}_+)$$

$$(M, a, b, a', b') \mapsto \left( n \mapsto [v \mapsto a \times v + b] \left( [v \mapsto a' \times v + b']^{(n)}(M) \right) \right)$$

- Intended to be a classical “non-relational” value domain (for filters, floats,...):  
attach a function bound on each variable, parametric in the value of a local loop counter.
- Coordinate-wise  $\sqsubseteq$ ,  $\sqcup$ , addition, ...
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Shift-friendly exponentials: **Stirling numbers of the 2<sup>nd</sup> kind** [Kahn&Hoffmann20].

- Replace basis  $1, 2^n, 3^n, \dots$  by  $\{1^n\}, \{2^n\}, \{3^n\}, \{4^n\}, \dots$

$$\left\{ \begin{matrix} n \\ k \end{matrix} \right\} \sim_{n \rightarrow \infty} \frac{k^n}{k!} \qquad \left\{ \begin{matrix} n+1 \\ k \end{matrix} \right\} = k \left\{ \begin{matrix} n \\ k \end{matrix} \right\} + \left\{ \begin{matrix} n \\ k-1 \end{matrix} \right\}$$

- Extends well to multivariate settings.

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- Extends well to multivariate settings.

Going beyond — **exponential polynomials**.

- Replace basis  $n \mapsto b^n \cdot n^k$  by  $n \mapsto \{n\}_b \{n\}_k$
- $\llbracket \text{Pop} \rrbracket^\#$  still works well, and  $\llbracket \text{Push } c \rrbracket^\#$  can be synthesised as before.

$$\{n+1\}_b \{n+1\}_k = b \{n\}_b \{n\}_k + b \{n\}_b \{n\}_{k-1} + \{n\}_{b-1} \{n\}_k + \{n\}_{b-1} \{n\}_{k-1}$$

# Discussion: Transfer Function Synthesis, Widenings, ...

## Partial conclusion

For such **sequence domains** (useful for loops, equations, recursive functions, streams, ...)

- **Non-linearity** is natural and reasonably **easy** to achieve.
- Embrace **regularity of functions** vs arbitrary relations.
- Important enabler: **convexity** properties of  $\{b \in \mathfrak{B} \mid \bigwedge_n f(n) \leq b(n)\}$   
(vs  $\{\phi \in \mathfrak{C} \mid \bigwedge_p \phi(p)\}$ )? Could we actually reuse some of these ideas for relations?).

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## Widenings

We have not discussed widenings. They are crucial. Our idea is

- Use **stability**  $\nabla$  + **thresholds**, parameter-wise (within  $\mathfrak{B}$ ),
- On  $\mathcal{P}_\uparrow(\mathfrak{B})$ : bound the number of constraints, **drop/join above max cardinality**,
- Perhaps **default to “fixed constraints”**? “local fixpoints”.

→ Can we do better? Is this appropriate?

## Discussion: Transfer function synthesis

- Still, **design of transfer functions** for a new  $\mathfrak{B}$  is the **main bottleneck** in creation of such abstract domains (more burdensome than **other postfixpoint search methods**, e.g. **optimisation-based**).
- **Transfer function synthesis** may be the right tool.  
We were able to generate several automatically via **CAS + SMT**, and more still in interactive loops.
- Can we streamline this synthesis process?
- **Extra precision**: generate **transfers for small combinations of basic constructs**,  $\llbracket s_1 \circ s_2 \circ \dots \rrbracket^\#$  vs  $\{\llbracket s_i \rrbracket^\#\}$ .



## Abstract Domains of Functions

Introducing more complex features:  
Multiple variables, disjunctivity,  
non-trivial recursive structure...

# Multivariate functions

When the boundaries form a vector space  $\mathfrak{B} = \bigoplus_k \mathbb{R} \cdot f_k$  extended with top elements, as often before, it is easy to move on from  $\mathbb{N} \rightarrow \mathbb{R}$  to  $\mathbb{N}^d \rightarrow \mathbb{R}$ .

Do a **(tensor) product**, and do products of coordinate-wise orders (for parameters).

$$\mathfrak{B}^{\otimes d} := \bigotimes_{i < d} \bigoplus_k \mathbb{R} \cdot f_k^{(i)} = \bigoplus_{k_1, \dots, k_d} \mathbb{R} \cdot f_{k_1}^{(1)} \otimes \dots \otimes f_{k_d}^{(d)}$$

## Example

Monovariate polynomials  $\mathbb{R}[n]$  in the monomial basis  $P(n) = \sum_k a_k n^k$  simply give rise to multivariate polynomials  $\mathbb{R}[x, y] = (\mathbb{R}[n])^{\otimes 2}$  in the monomial basis  $P(x, y) = \sum a_{k_x, k_y} x^{k_x} y^{k_y}$ .

Transfer functions ( $\llbracket \Diamond \rrbracket^\sharp$ ,  $\llbracket \text{Cst } c \rrbracket^\sharp$ , etc.) are easily extended.

We can similarly define (and compose)  $\llbracket \text{Pop}_k \rrbracket^\sharp$  and  $\llbracket \text{Push}_k c \rrbracket^\sharp$  on each dimension,

e.g.,  $\text{Push}_y c : f(x, y, z) \rightsquigarrow \text{ite}(y = 0, c, f(x, y - 1, z))$ .

## Example

For polynomials,  $1 \cdot (x + 1)^2 yz = (x^2 + 2x + 1)yz = 1 \cdot x^2 yz + 2 \cdot xyz + 1 \cdot yz$ . Similarly, in the binomial basis,  $\binom{x+1}{k_x} \binom{y}{k_y} \binom{z}{k_z} = \binom{x}{k_x} \binom{y}{k_y} \binom{z}{k_z} + \binom{x}{k_x-1} \binom{y}{k_y} \binom{z}{k_z}$ , so more generally

$$\llbracket \text{Pop}_x \rrbracket^\sharp \left( \sum a_{(k_x, k_y, k_z)} \binom{x}{k_x} \binom{y}{k_y} \binom{z}{k_z} \right) = \left\{ \sum (a_{(k_x, k_y, k_z)} + a_{(k_x+1, k_y, k_z)}) \binom{x}{k_x} \binom{y}{k_y} \binom{z}{k_z} \right\}.$$

# Piecewise functions — more complex control flows



## Example

```
while (0 < i && i < n) {  
  if (b) i++;  
  else i--;  
}
```

Simple loop (from [AFloresPhd17])

$$\Phi(f)(i, n, b) = \begin{cases} 0 & \text{if } i = 0 & (\mathcal{D}_1) \\ 0 & \text{if } i \geq n & (\mathcal{D}_2) \\ 1 + f(i + 1, n, b) & \text{if } 0 < i < n \wedge b \geq 1 & (\mathcal{D}_3) \\ 1 + f(i - 1, n, b) & \text{if } 0 < i < n \wedge b = 0 & (\mathcal{D}_4) \end{cases}$$

Corresponding cost equation

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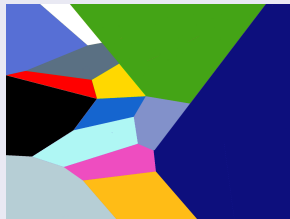
## Piecewise function domains (closely related to BDT domains [Urban&Mine14])

Sets of  **$\mathfrak{B}$ -bounds by  $\mathcal{C}$ -cases**, where  $\mathcal{C}$  is a constraint domains (e.g. polyhedra, etc.).

$$\mathbb{P}(\mathfrak{B}, \mathcal{C}) \ni \{(F_1, c_1), \dots, (F_k, c_k)\}, \text{ s.t.}$$

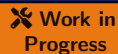
- $F_i \in \mathcal{P}_\uparrow(\mathfrak{B})$ ,
- the  $c_i \in \mathcal{C}$  partition the input space  $\mathcal{D}$ .

→ Instead of one conjunction of bounds, piecewise conjunction of bounds.



Key primitive: ability to **merge cases**, i.e.  
“Compute best  $\mathfrak{B}$ -approximation of  $\mathfrak{B}$ -cases by  $\mathcal{C}$ -parts.”

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$$\perp^\# = \begin{cases} \{0\} & (\mathcal{D}_1) \\ \{0\} & (\mathcal{D}_2) \\ \{0\} & (\mathcal{D}_3) \\ \{0\} & (\mathcal{D}_4) \end{cases}$$

## Domains of piecewise functions

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$$(\Phi^\#)^{(2)}(\perp^\#) = ?$$

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}
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Simple loop (from [AFloresPhd17])

$$\Phi(f)(i, n, b) = \begin{cases} 0 & \text{if } i = 0 & (\mathcal{D}_1) \\ 0 & \text{if } i \geq n & (\mathcal{D}_2) \\ 1 + f(i + 1, n, b) & \text{if } 0 < i < n \wedge b \geq 1 & (\mathcal{D}_3) \\ 1 + f(i - 1, n, b) & \text{if } 0 < i < n \wedge b = 0 & (\mathcal{D}_4) \end{cases}$$

Corresponding cost equation

$$(\Phi^\#)(\perp^\#) = \begin{cases} \{0\} & (\mathcal{D}_1) \\ \{0\} & (\mathcal{D}_2) \\ \{1\} & (\mathcal{D}_3) \\ \{1\} & (\mathcal{D}_4) \end{cases}$$

$$(\Phi^\#)^{(2)}(\perp^\#) = ?$$

$$(\Phi \circ \Phi^\#)(\perp^\#) = \begin{cases} 0 & (\mathcal{D}_1) \\ 0 & (\mathcal{D}_2) \\ \begin{cases} 2 & \text{if } (i, n, b) \in \mathcal{D}_3 \wedge (i + 1, n, b) \in \mathcal{D}_3 \\ 1 & \text{if } (i, n, b) \in \mathcal{D}_3 \wedge (i + 1, n, b) \in \mathcal{D}_2 \end{cases} \\ \begin{cases} 1 & \text{if } (i, n, b) \in \mathcal{D}_4 \wedge (i - 1, n, b) \in \mathcal{D}_1 \\ 2 & \text{if } (i, n, b) \in \mathcal{D}_4 \wedge (i - 1, n, b) \in \mathcal{D}_4 \end{cases} \end{cases}$$

## Domains of piecewise functions

$$\mathbb{P}(\mathfrak{B}, \mathcal{C}) \ni \{(F_1, c_1), \dots, (F_k, c_k)\},$$

- $F_i \in \mathcal{P}_\uparrow(\mathfrak{B})$ ,
- the  $c_i \in \mathcal{C}$  partition  $\mathcal{D}$ .

$$\rightarrow \vec{n} \in \gamma(c_i) \Rightarrow \bigwedge_j f(\vec{n}) \leq f_{i,j}^\#(\vec{n})$$

Key primitive:

“Compute best  $\mathfrak{B}$ -approximation of  $\mathfrak{B}$ -cases by  $\mathcal{C}$ -parts.”

→ Refine at **interfaces between subdomains**, and **merge subcases**.



# Piecewise functions — more complex control flows



## Example

```
while (0 < i && i < n) {
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$$(\Phi^\#)^{(2)}(\perp^\#) = \begin{cases} \{0\} & (\mathcal{D}_1) \\ \{0\} & (\mathcal{D}_2) \\ \{n - i\} & (\mathcal{D}_3) \\ \{i\} & (\mathcal{D}_4) \end{cases}$$

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$$\leftrightarrow f_{sol}(i, n, b) !$$

## Domains of piecewise functions

- $\mathbb{P}(\mathfrak{B}, \mathcal{C}) \ni \{(F_1, c_1), \dots, (F_k, c_k)\}$ ,
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Key primitive:

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# Continuous systems: example of ODE



Abstract functions become particularly interesting in cases where we cannot (or do not want to) compute input/output samples, e.g. for **continuous systems**, with a function  $f : \mathbb{R}_+ \rightarrow \overline{\mathbb{R}}$  defined as the solution of a **differential equation**.  
 → Towards  $\mathfrak{B} \in \mathcal{P}(\mathbb{R}_+ \rightarrow \overline{\mathbb{R}})$ ?

$$v : \mathbb{R}_+ \rightarrow \mathbb{R}, \quad v(0) = v_0, \quad \dot{v} = -\alpha \cdot v^2 - \beta \cdot v + \gamma$$

**Non-standard analysis** with infinitesimal  $\epsilon$ , within **hyperreals**  ${}^*\mathbb{R}$ ?

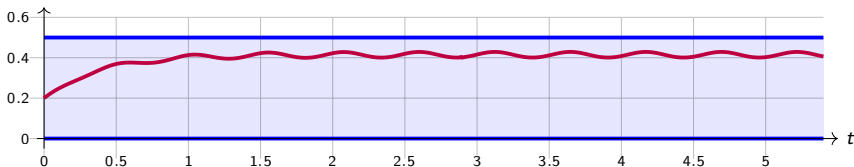
Or more simply a family of Euler schemas.

$$\Phi \in \text{End}({}^*\mathbb{R}_+ \rightarrow \mathcal{I}({}^*\mathbb{R}), \dot{\subseteq}_{\mathcal{I}})$$

$$:= f \mapsto t \mapsto \begin{cases} [v_0, v_0] & \text{if } t < \epsilon \\ f(t - \epsilon) - \alpha \cdot \epsilon \cdot f(t - \epsilon) \cdot \mathcal{I} f(t - \epsilon) - \beta \cdot \epsilon \cdot f(t - \epsilon) + \epsilon \cdot \gamma & \text{if } t \geq \epsilon \end{cases}$$

For  $\hat{f} : t \mapsto [0, M]$ ,  $\Phi \hat{f} \dot{\subseteq}_{\mathcal{I}} \hat{f}$  whenever  $x_0 \in [0, M]$ ,  $\alpha, \beta, \gamma \geq 0$  and  $M \geq \gamma/\beta$ .

→ Can also be applied with time-dependent  $\alpha(t), \beta(t), \gamma(t)$ , which is typically not solvable.



Example with  $\alpha(t) \in [0, 2]$ ,  $\beta = 2$ ,  $\gamma = 1$ ,  $v_0 = 0.2$ .

# Thank you!

Questions?

# Order theory framework of equations [SAS24]

Equations  $\leftrightarrow$  Operators

Solutions  $\leftrightarrow$  Fixpoints

Bounds  $\leftarrow$  Pre/Postfixp

## Example

Search  $f : \mathbb{N}^2 \rightarrow \mathbb{R}$  such that

$$f(n, c) = \begin{cases} f(n-1, 0) + n + 300 & \text{if } n > 0 \text{ and } c \geq 100, \\ f(n-1, c+1) + n & \text{if } n > 0 \text{ and } c < 100, \\ c & \text{if } n = 0, \end{cases}$$

$\rightarrow$

## Example

Search  $f : \mathbb{N}^2 \rightarrow \mathbb{R}$  such that  $f = \Phi f$ , i.e.  $f \in \text{Fixp}(\Phi)$ , where

$$\Phi : (\mathbb{N}^2 \rightarrow \mathbb{R}) \rightarrow (\mathbb{N}^2 \rightarrow \mathbb{R})$$

$$f \mapsto (n, c) \mapsto \begin{cases} f(n-1, 0) + n + 300 & \text{if } n > 0 \text{ and } c \geq 100, \\ f(n-1, c+1) + n & \text{if } n > 0 \text{ and } c < 100, \\ c & \text{if } n = 0. \end{cases}$$

For a complete lattice, order  $\mathcal{D} \rightarrow \mathbb{R}$  pointwise, and extend to  $\overline{\mathbb{R}} := \mathbb{R} \cup \{\pm\infty\}$

## Theorem

Let  $\Phi : (\mathcal{D} \rightarrow \overline{\mathbb{R}}) \rightarrow (\mathcal{D} \rightarrow \overline{\mathbb{R}})$  be a **monotone equation**.

- If  $f \in \text{Postfp}(\Phi)$ , i.e.  $\Phi f \leq f$ , then  $\text{lfp } \Phi \leq f$ .
- If  $f \in \text{Prefp}(\Phi)$ , i.e.  $f \leq \Phi f$ , then  $f \leq \text{gfp } \Phi$ .

Insight: cost equations are typically monotone for this pointwise order,  
(and terminating  $\rightsquigarrow \text{lfp } \Phi = \text{gfp } \Phi$ ).