

LRA Interpolants from No Man's Land

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University of Lugano, Switzerland

THE LEGEND OF GNOME ANN

TIME AND TIDE WAIT
FOR GNOME ANN.



THE WICKED FLEE WHEN
GNOME ANN PURSUETH.
—PROVERBS 28:1



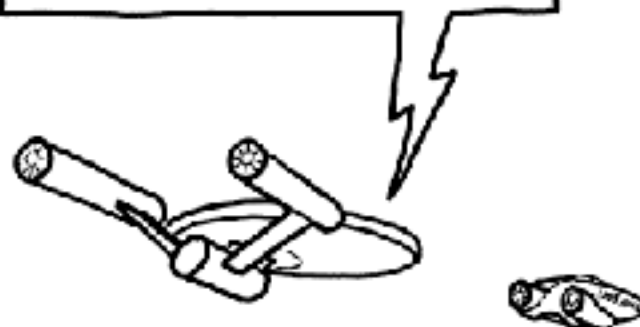
WHAT THEREFORE GOD HATH
JOINED TOGETHER, LET
GNOME ANN PUT ASUNDER.
—MARK 10:9



TIME RIPENS ALL THINGS;
GNOME ANN IS BORN WISE.
—MIGUEL DE CERVANTES



OUR MISSION: TO BOLDLY
GO WHERE GNOME ANN
HAS GONE BEFORE.



FOOL! NO MAN CAN KILL ME.
I AM GNOME ANN!



Motivation

The goal: Finding the right proof

The tool: Make interpolation on LRA more flexible

The application: LRA for abstractions in software model checking

The keywords: SMT solving, function summaries,
labeled interpolation systems

Interpolants

Given two formulas A and B such that

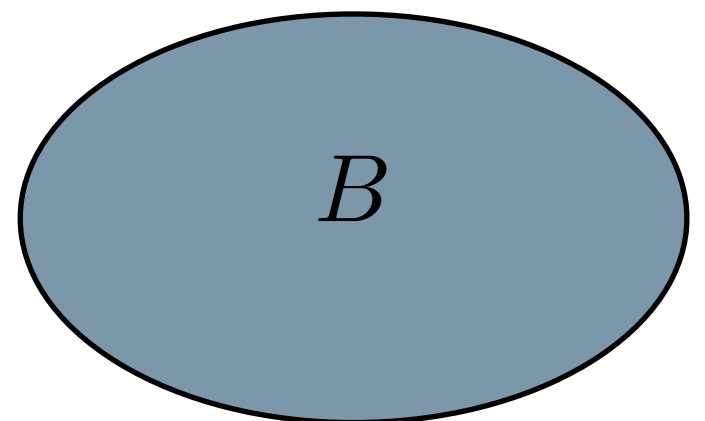
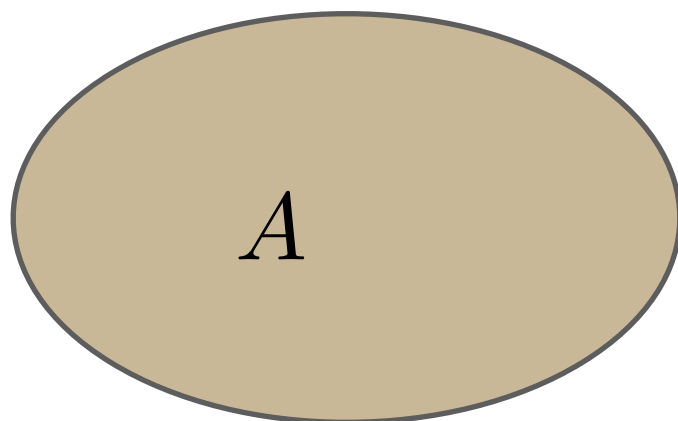
$$A \wedge B \rightarrow \perp$$

an **interpolant** is a formula I such that

$$\text{Vars}(I) \subseteq \text{Vars}(A) \cap \text{Vars}(B)$$

$$A \rightarrow I$$

$$I \wedge B \rightarrow \perp$$



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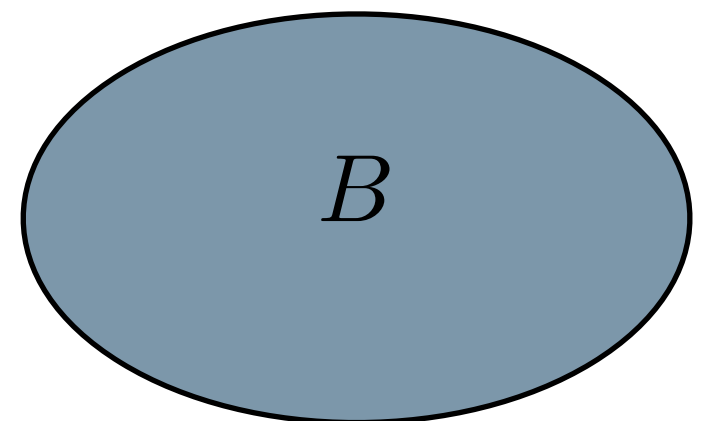
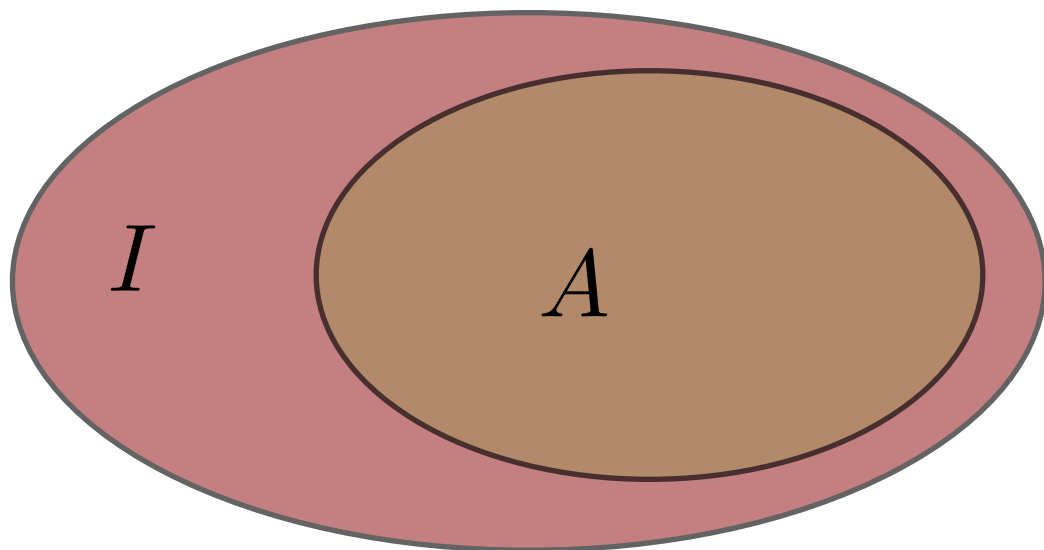
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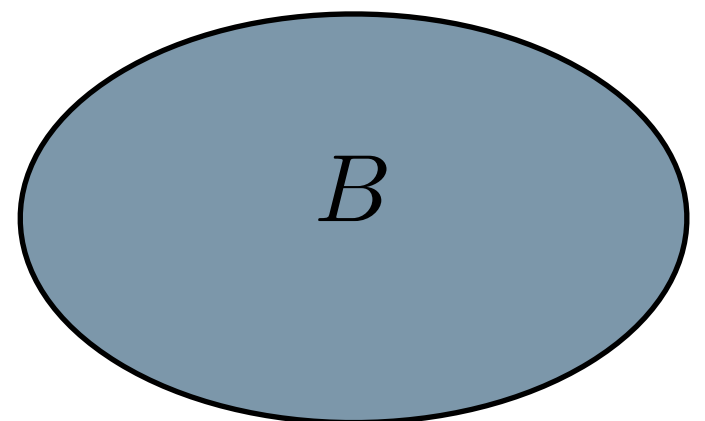
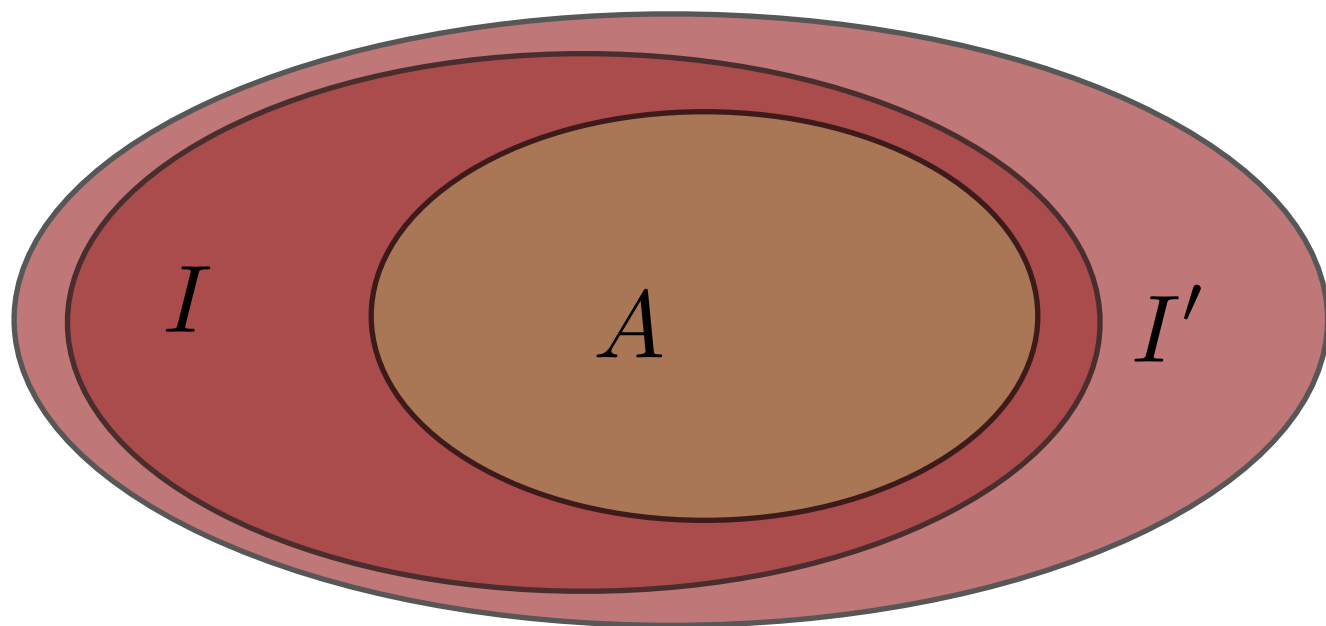
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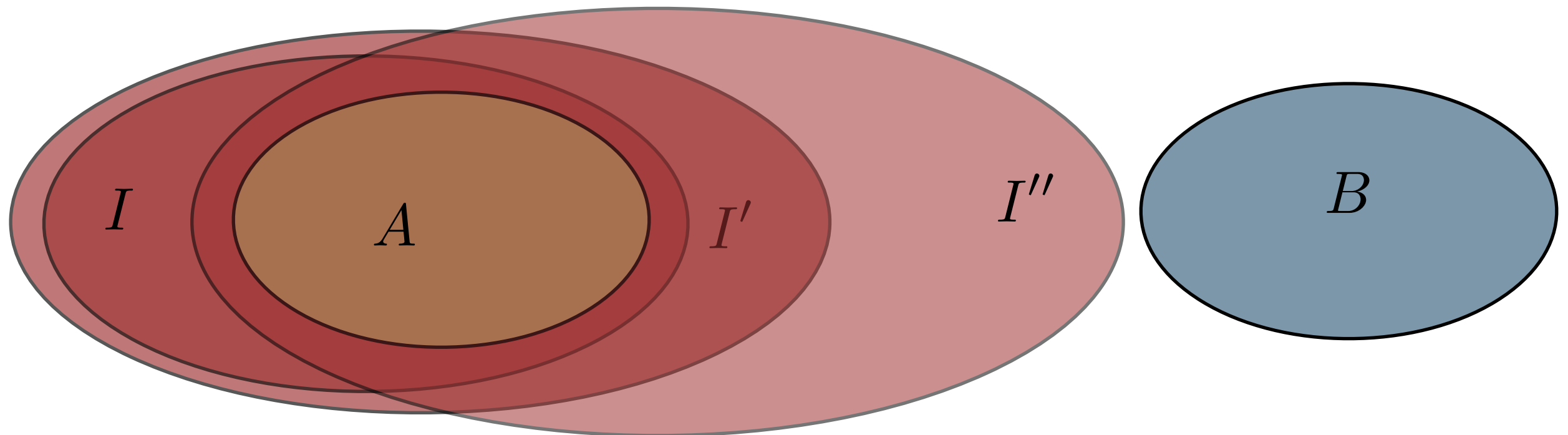
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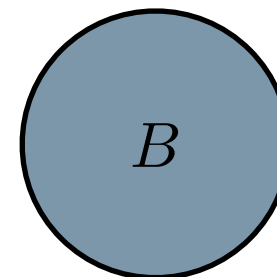
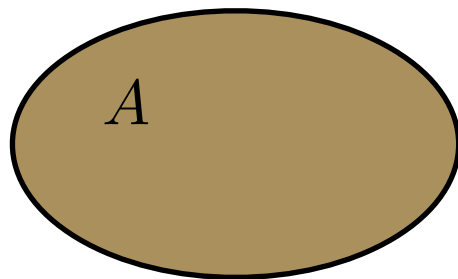
Interpolation in Proofs

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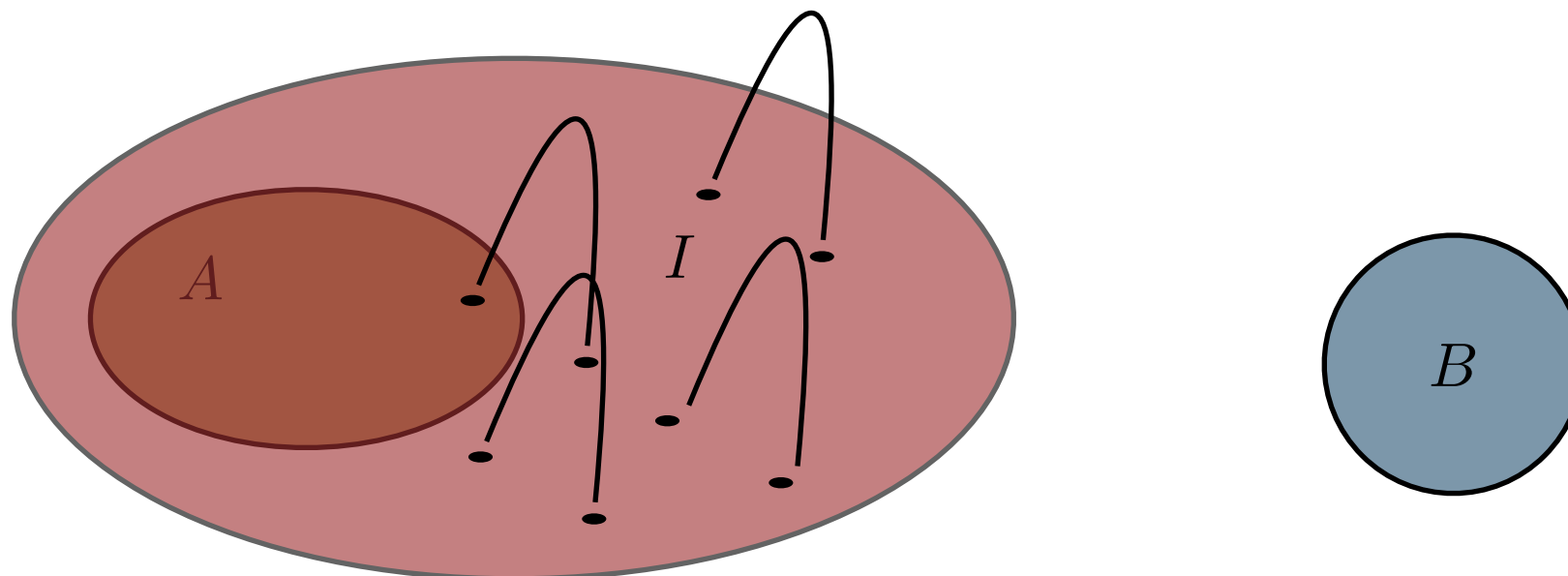
$$\underbrace{S(x_0)}_A \wedge \underbrace{T(x_0, x_1) \wedge \dots \wedge T(x_{k-1}, x_k) \wedge Err(x_k)}_B$$



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$$\underbrace{S(x_0)}_A \wedge \underbrace{T(x_0, x_1) \wedge \dots \wedge T(x_{k-1}, x_k) \wedge Err(x_k)}_B$$
$$I(x) \wedge T(x, x') \rightarrow I(x')$$



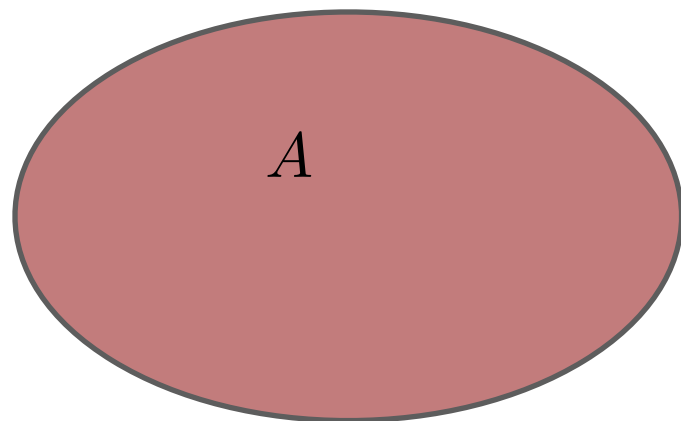
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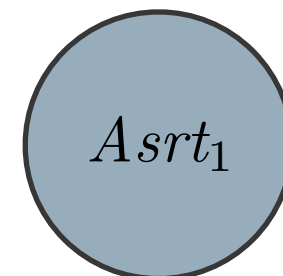
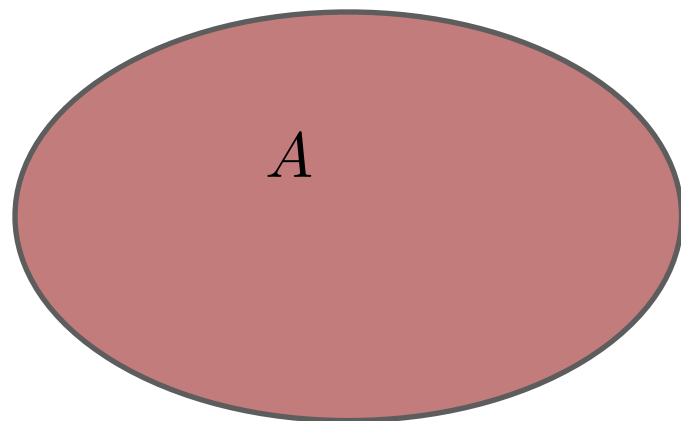
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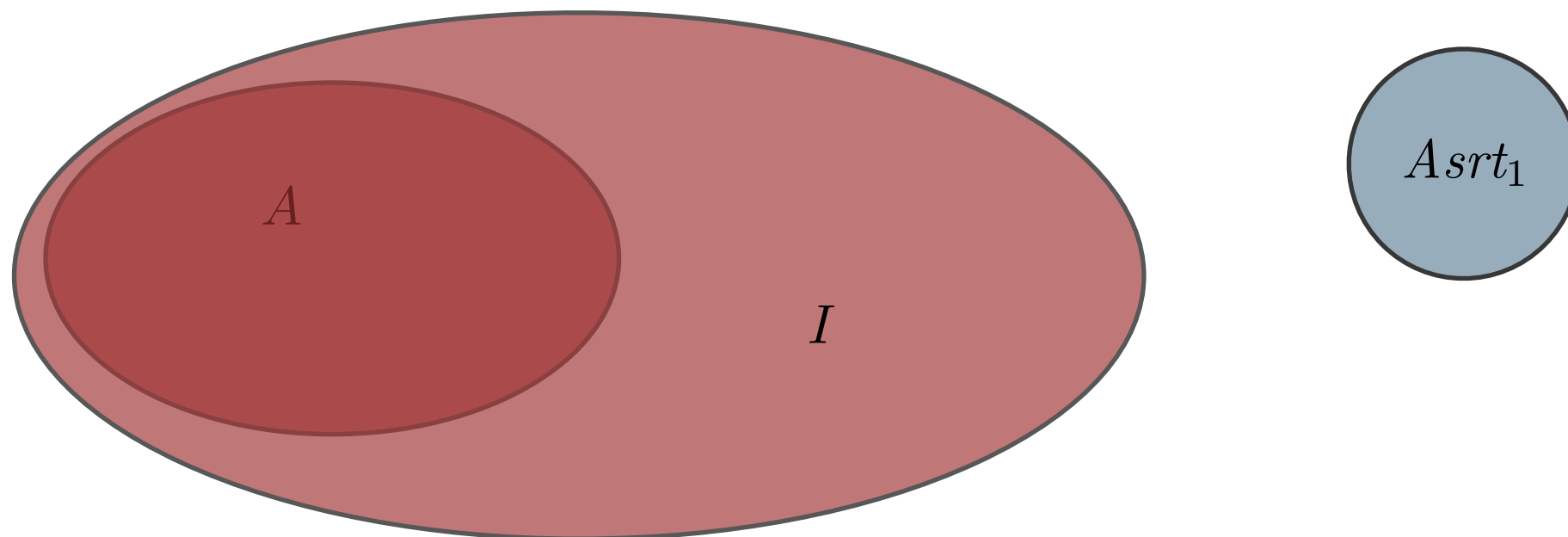
1. Construct a BMC instance of the program
2. Check the first assertion against the BMC instance



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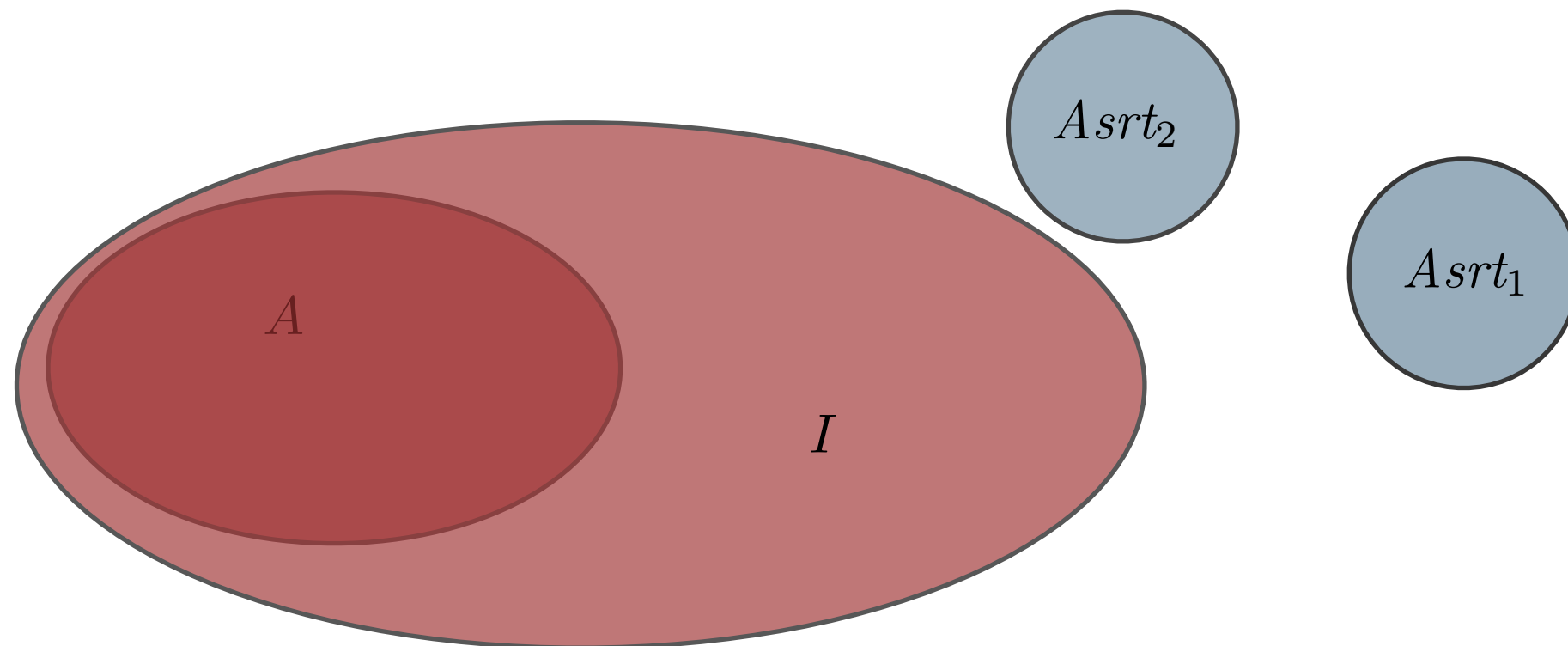
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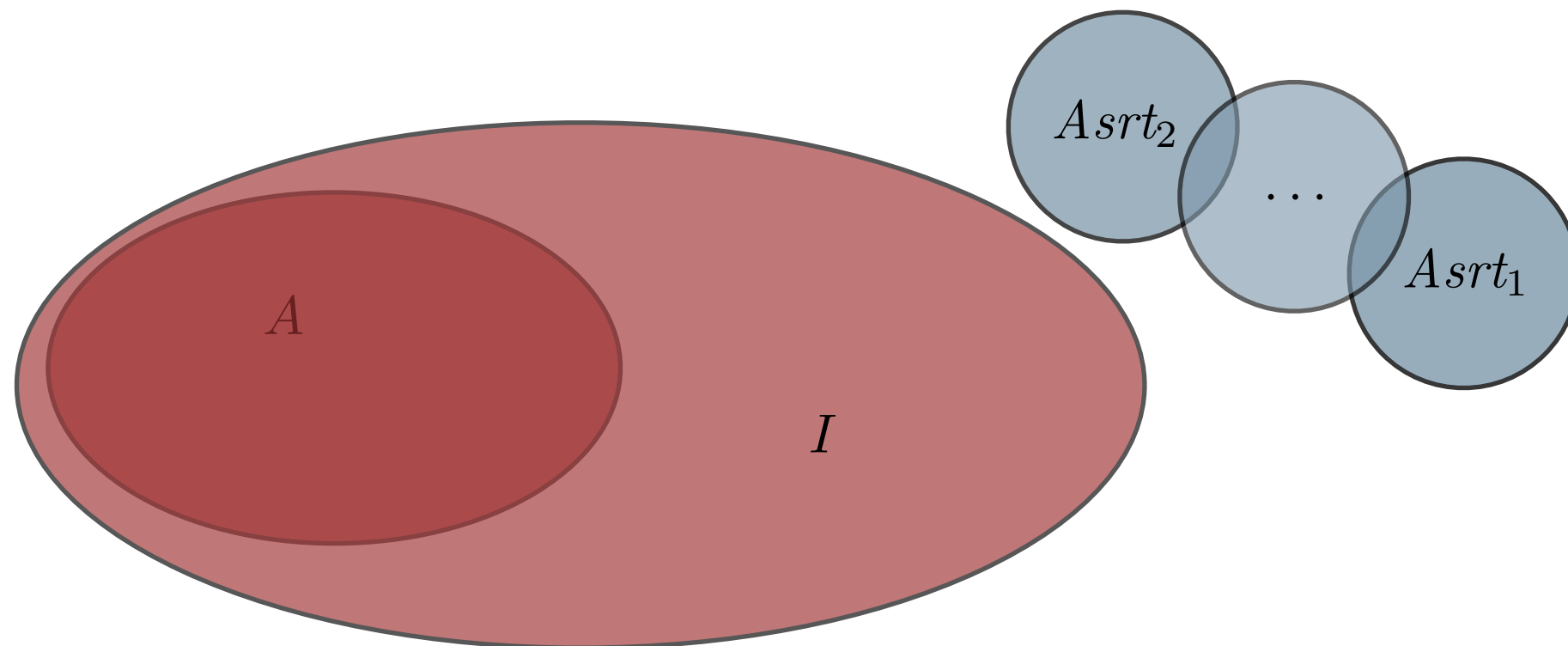
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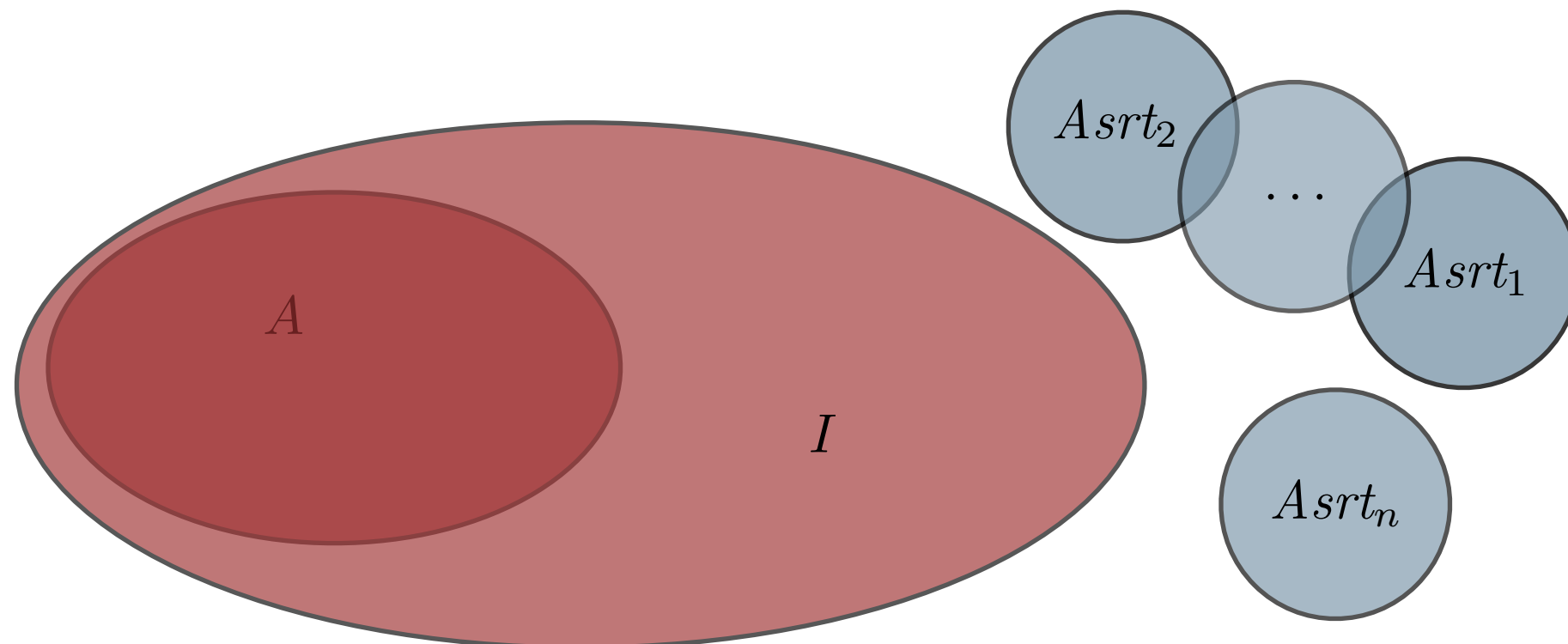
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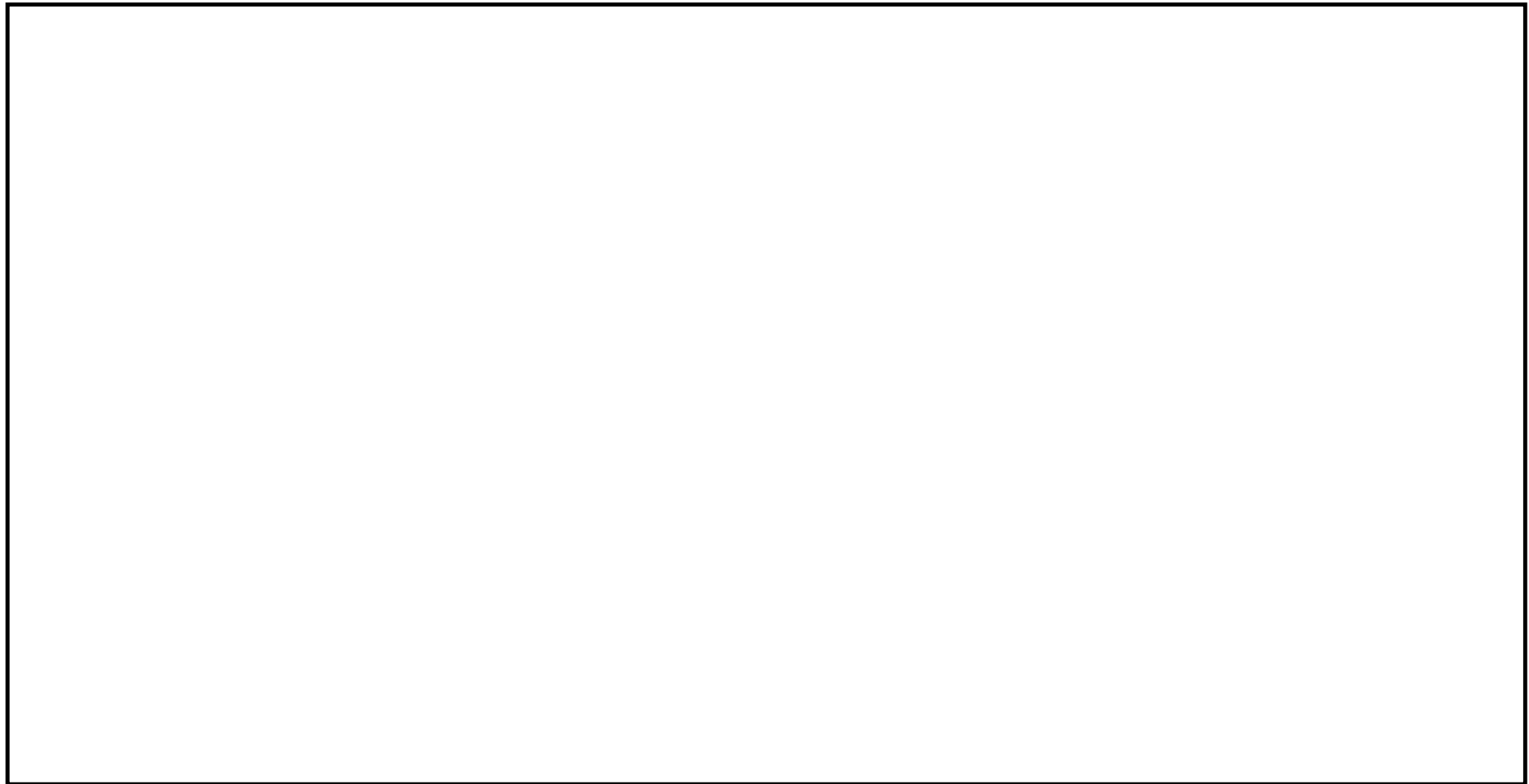
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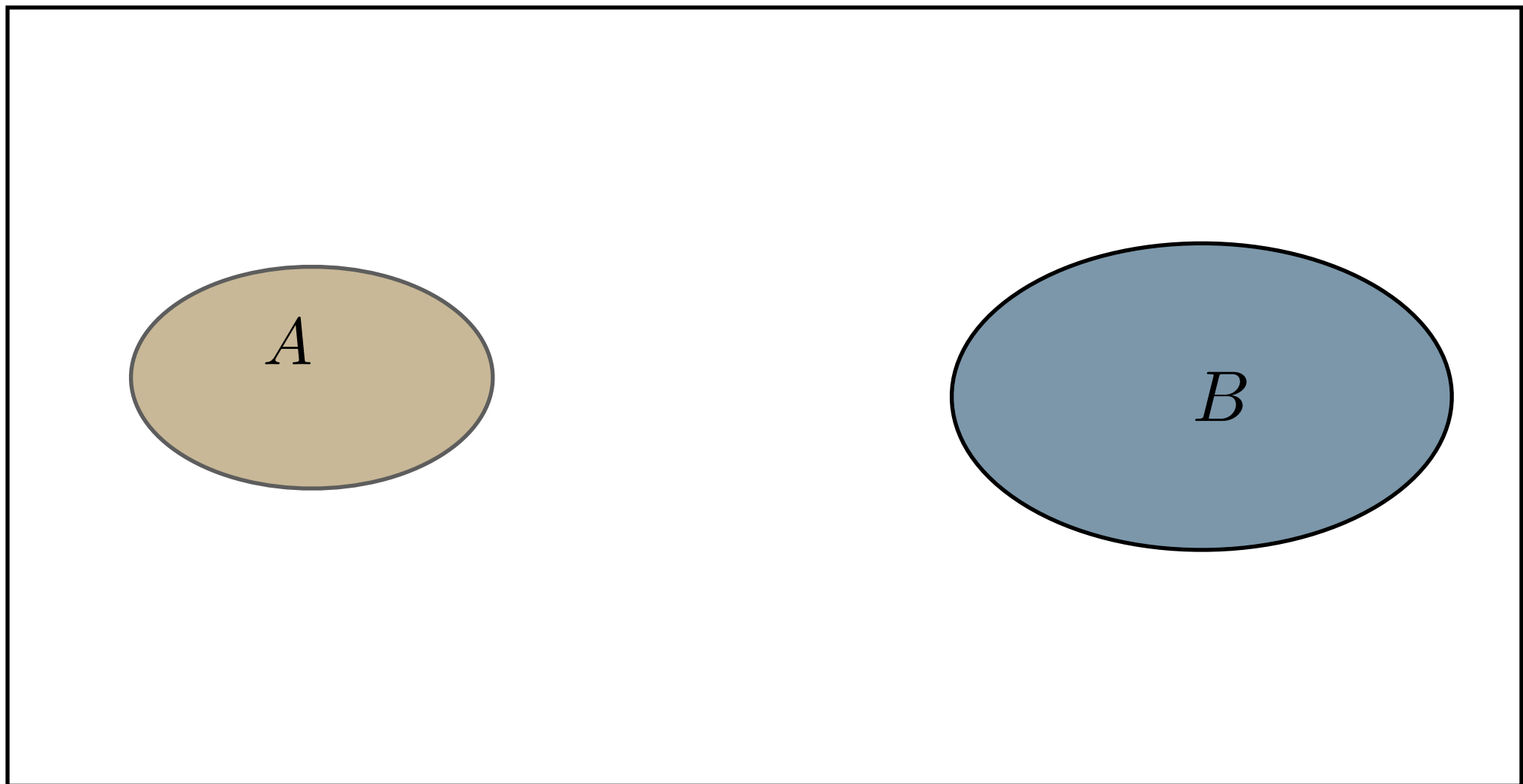
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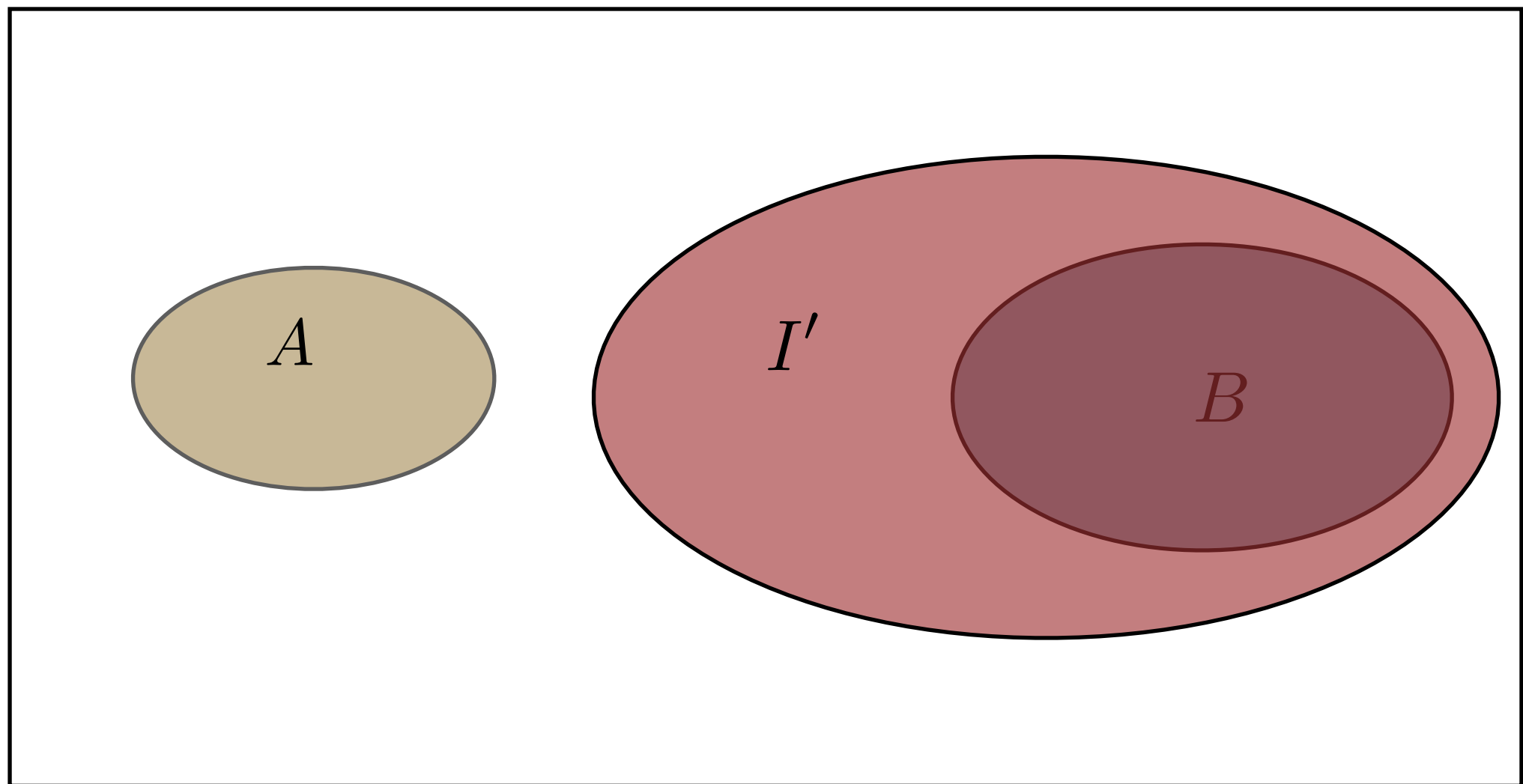
Duality of Interpolants



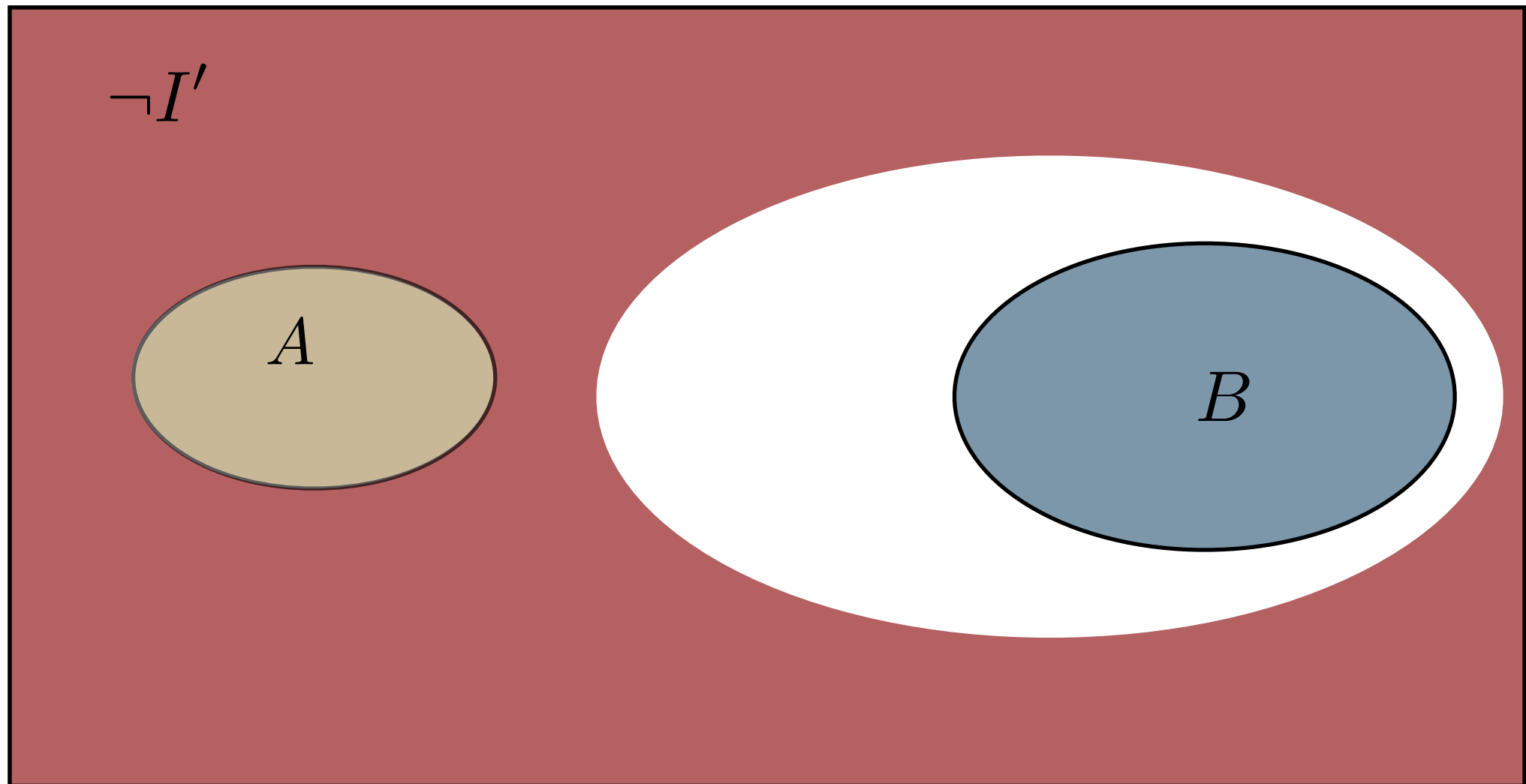
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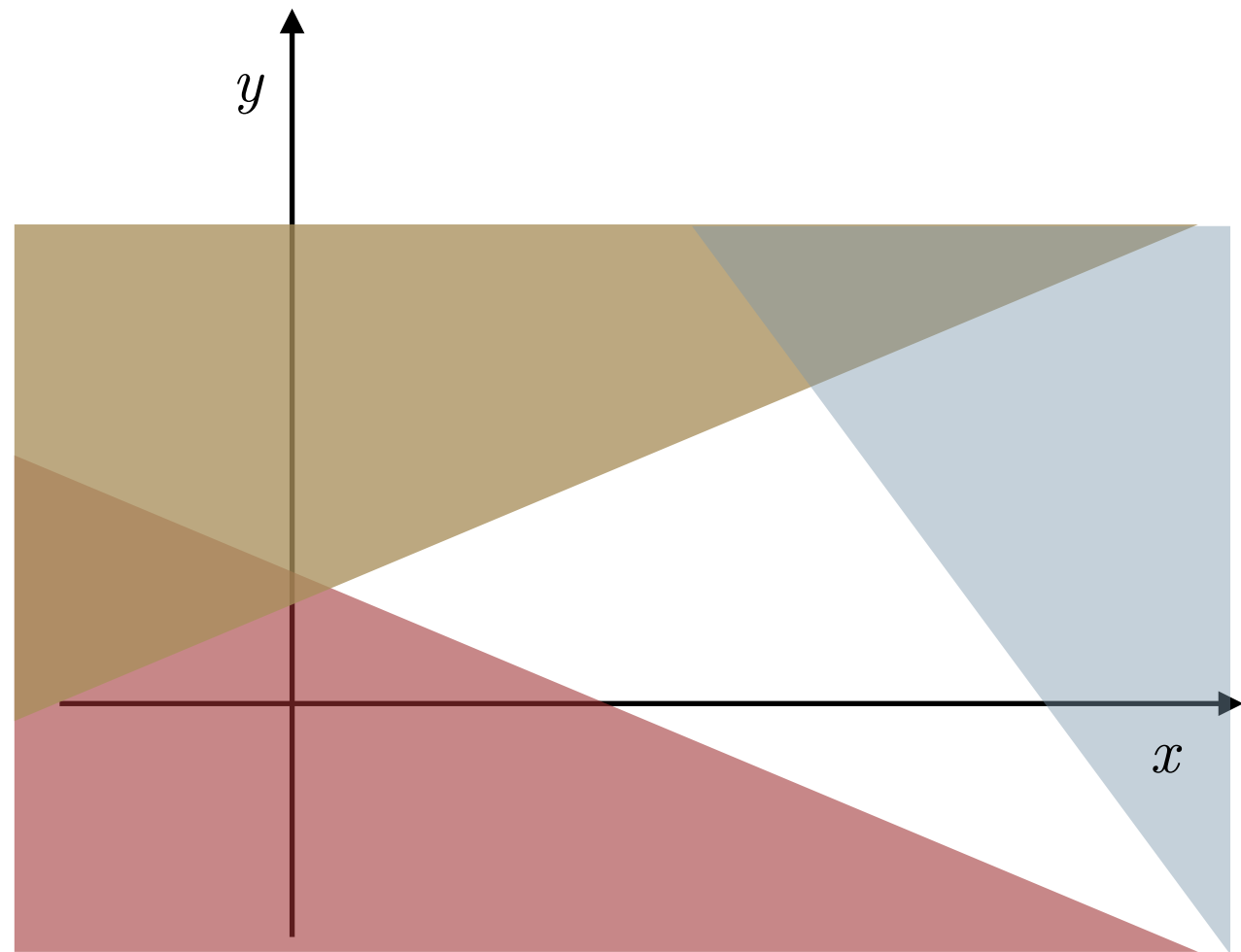
What is LRA

Given a set of linear inequalities over real-valued variables,
determine if there are values for the variables that satisfy all the inequalities

$$0 \leq 0.5x + 3y - 2z$$

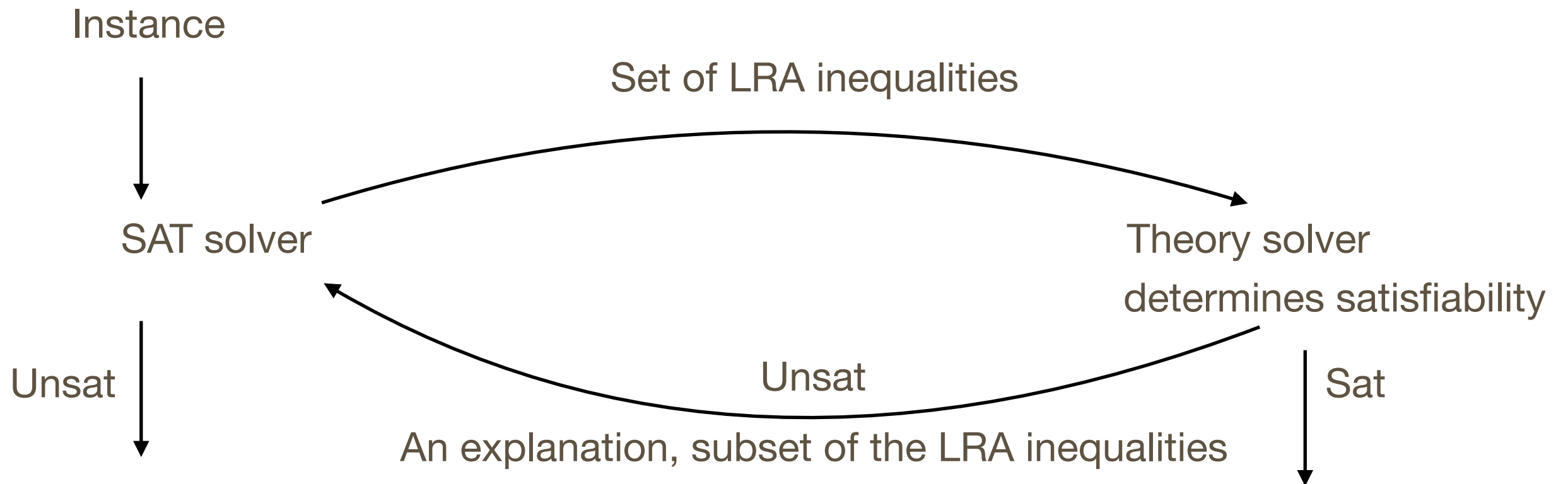
$$5 > y$$

~~$$x^2 + 2xy + y^2 > 1$$~~



In 2 dimensions: determine whether half planes have a non-empty intersection

Solving LRA in SMT



The theory solver for LRA is based on the Simplex algorithm

Simplex in SMT

A pre-processing step:

- All inequalities are written so that **left side** is a constant and **right side** a linear expression

We end up with two types of entities:

- Bounds on variables
- Bounds on sums of the variables

The idea is to repeatedly adjust variable values to satisfy bounds on the sums, and change the role of the variables and the sums.

$$0 \leq x + 6y - 4z$$

$$0.3 \geq x + 2y$$

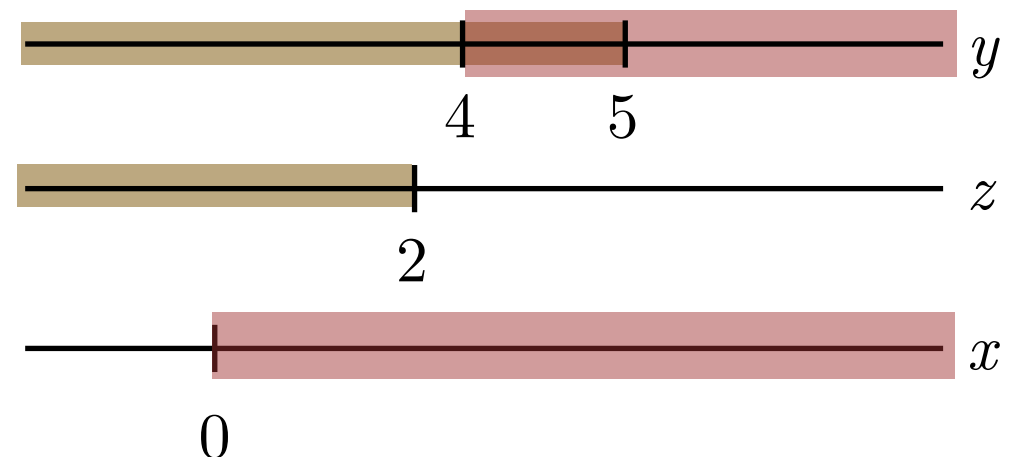
$$1 < x - 2y + 3z$$

$$5 > y$$

$$4 < y$$

$$2 > z$$

$$0 \geq x$$

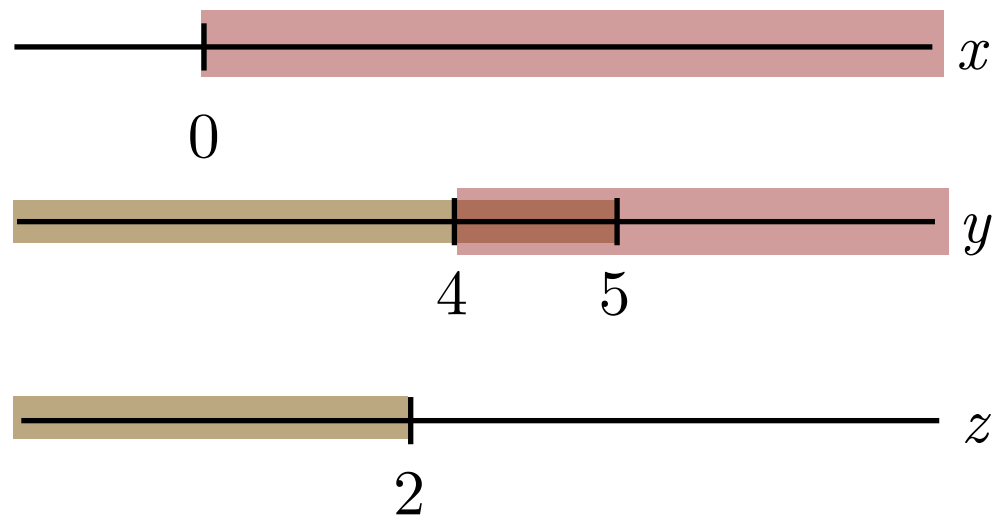


Simplex Example

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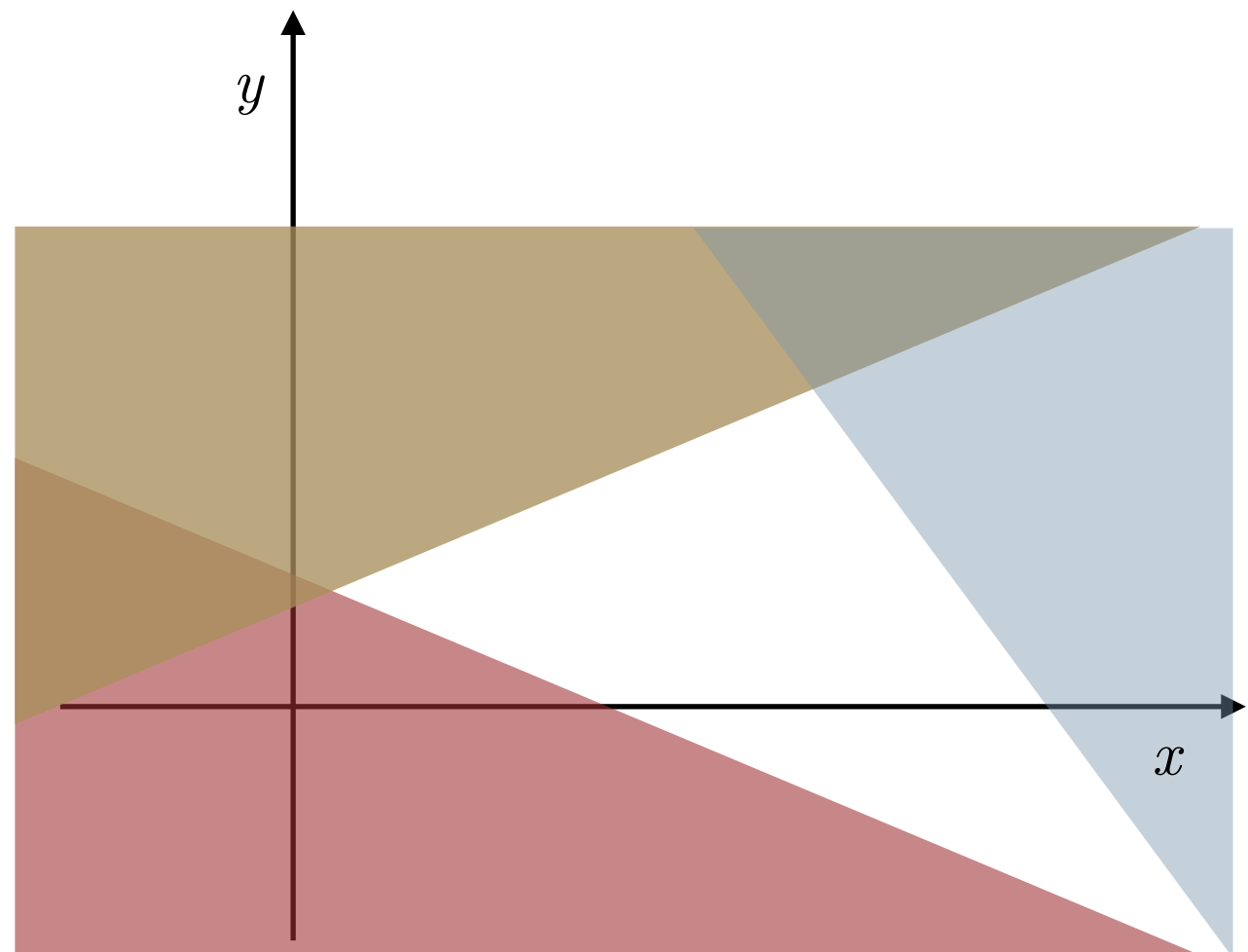
 \leq 

 \geq 

 $<$ 

 x

 y





Simplex Example

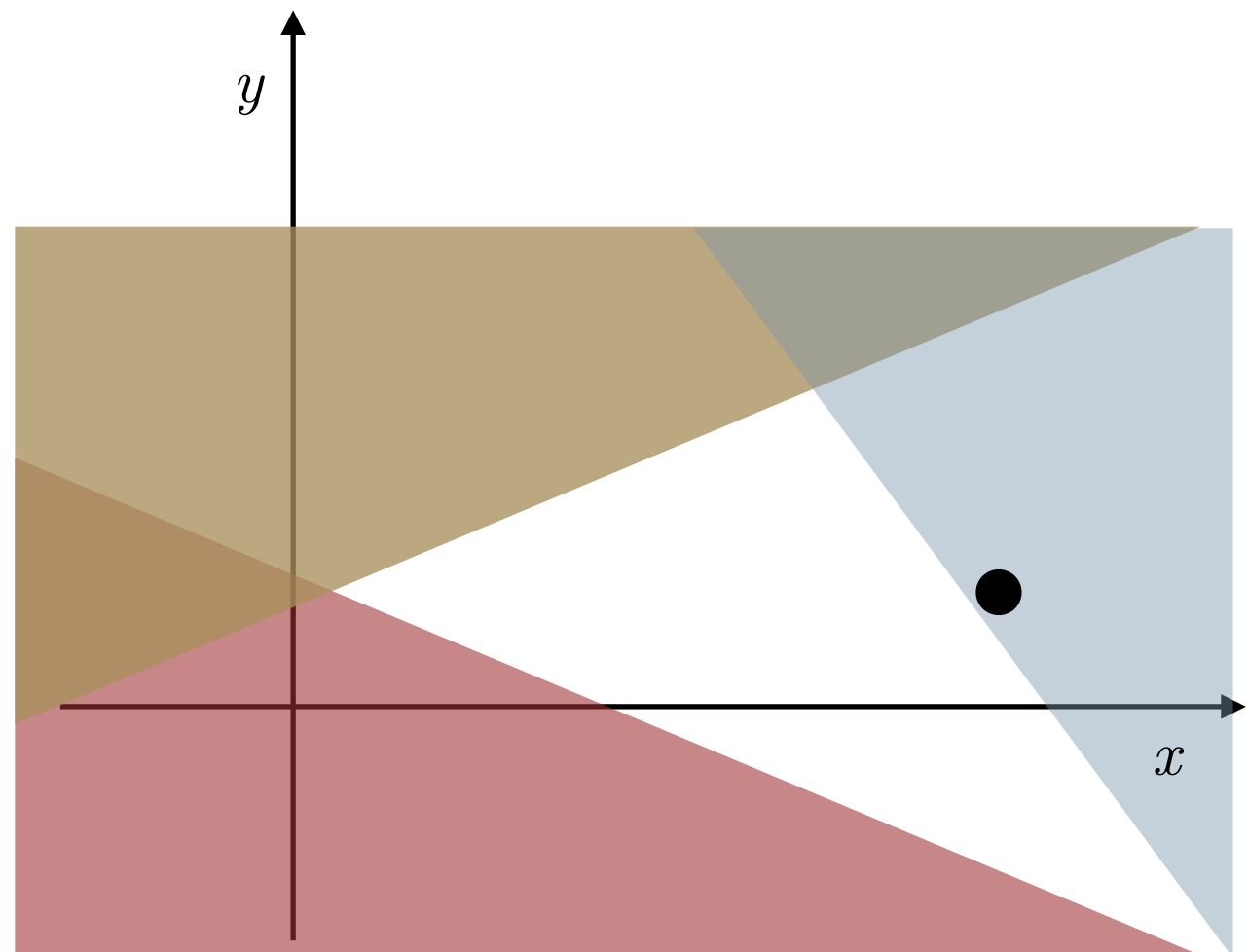
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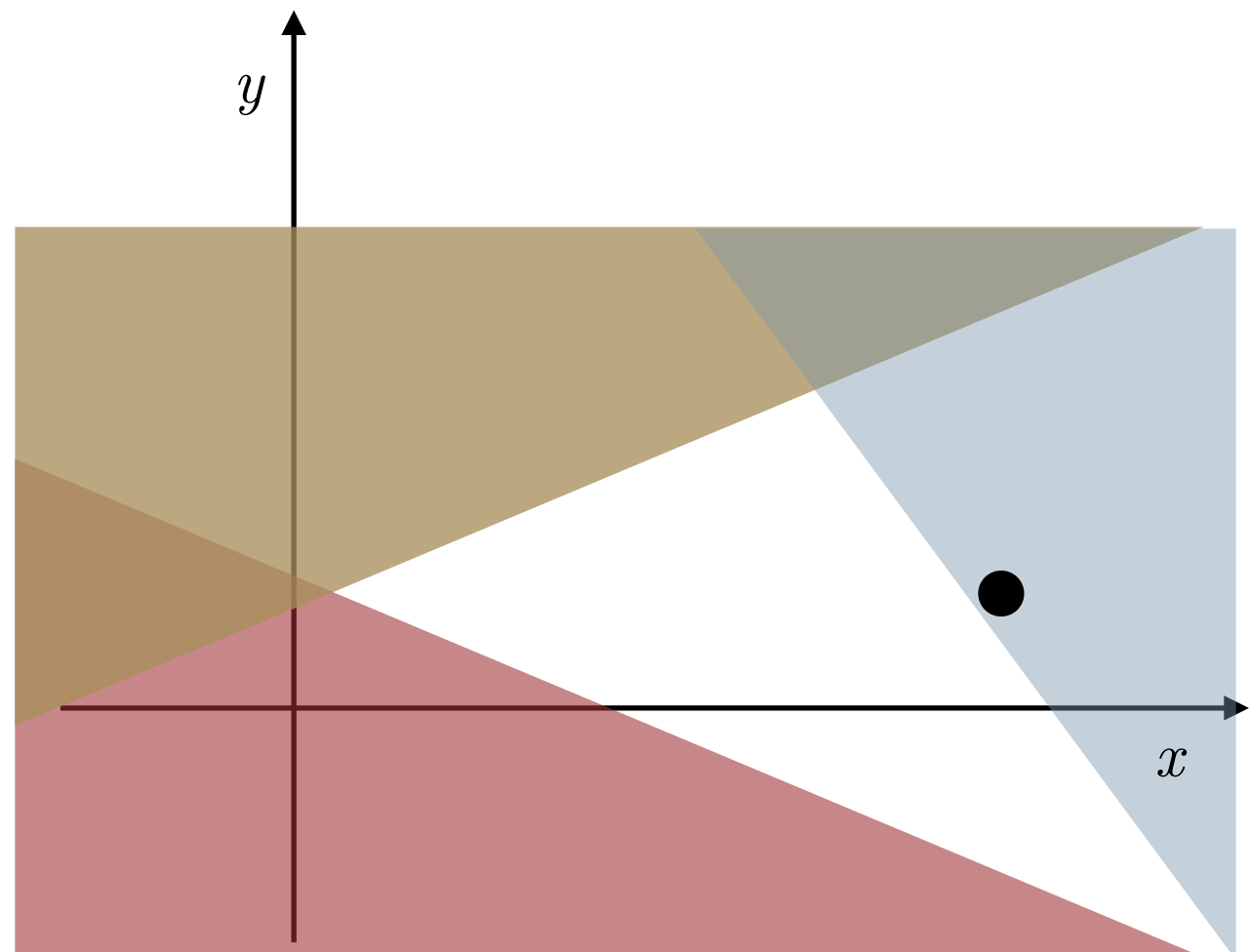
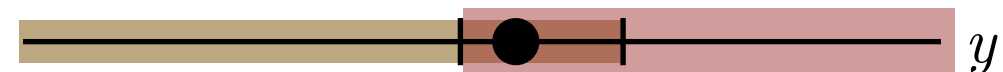
 $<$ 

 x

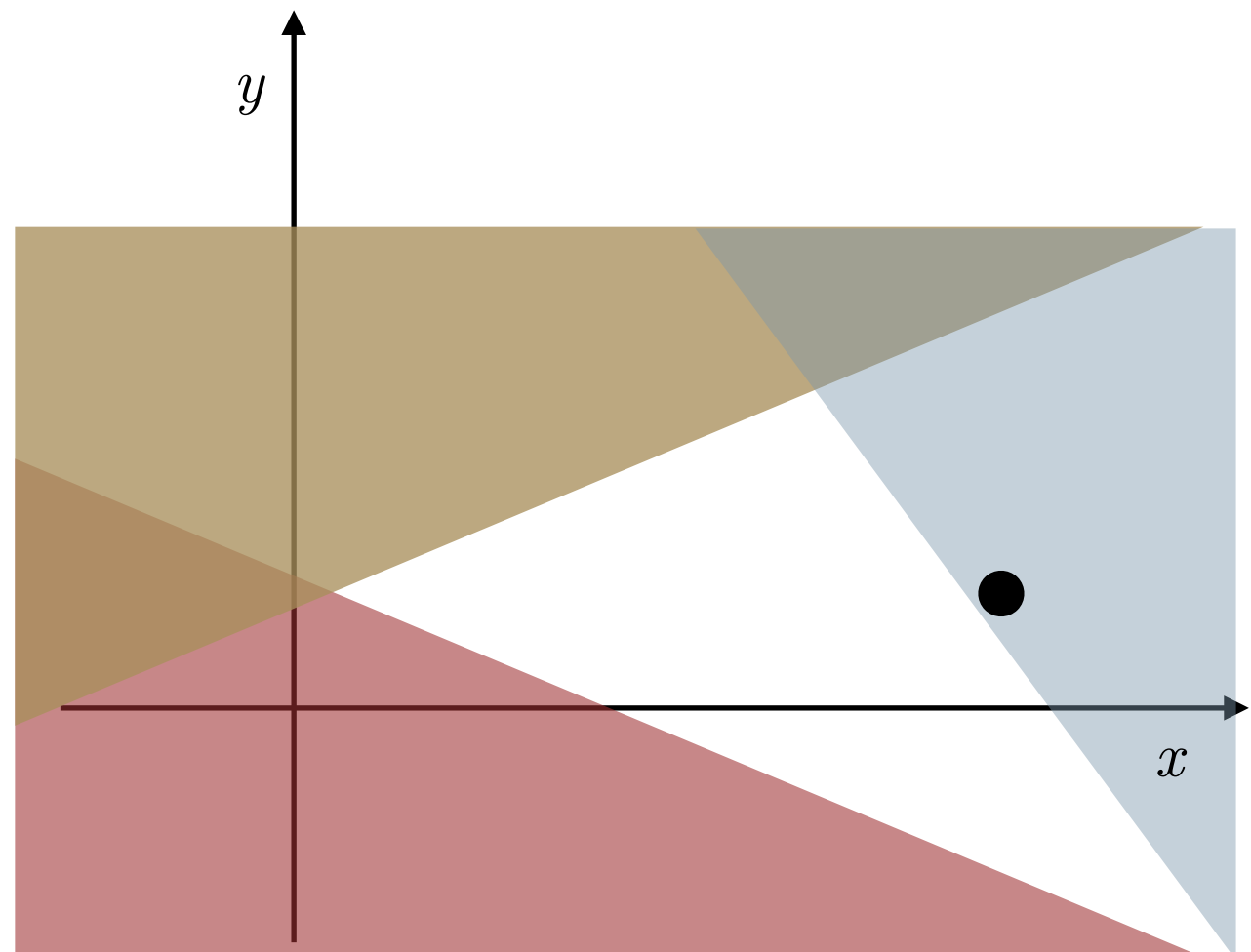
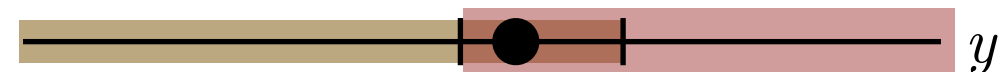
  y



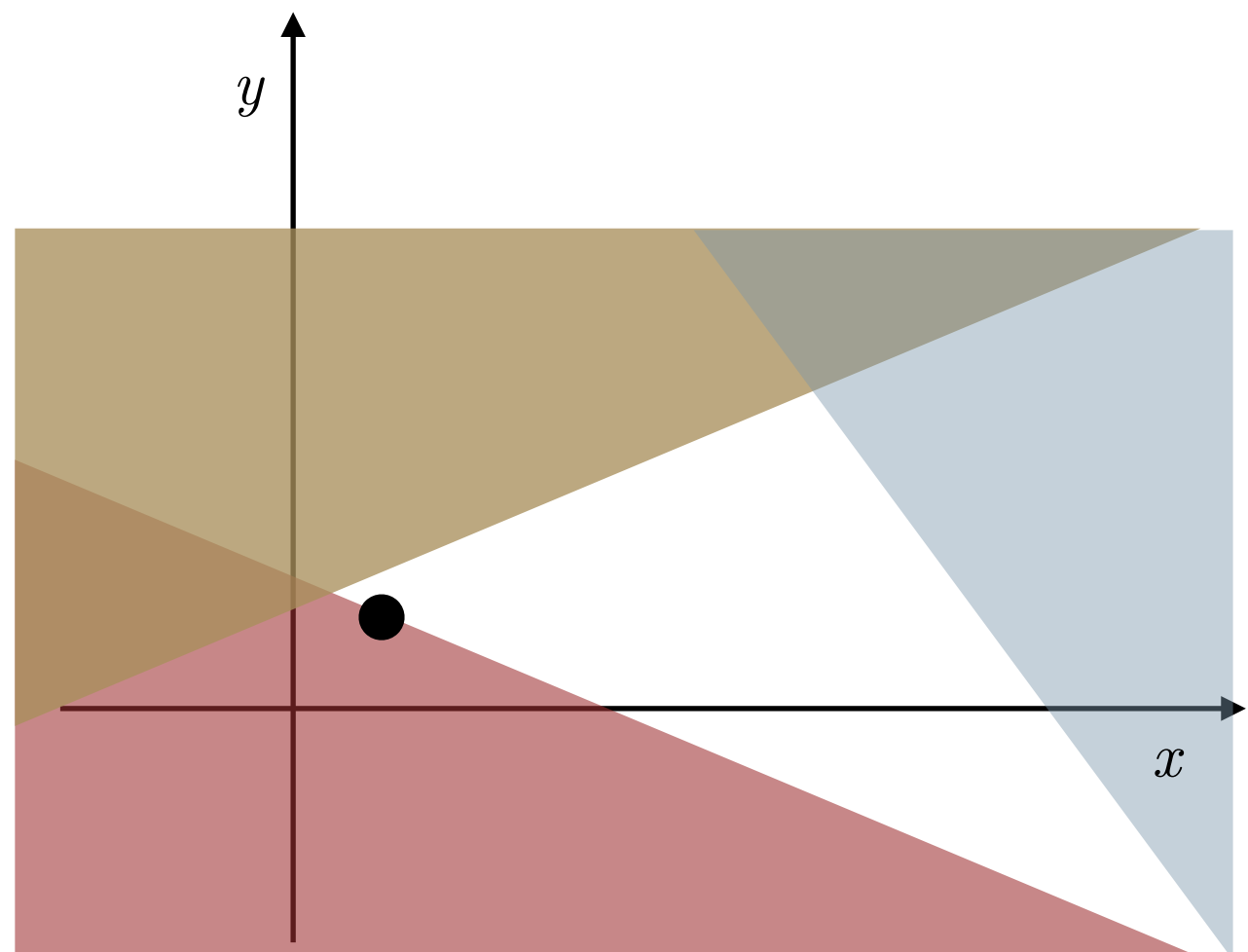
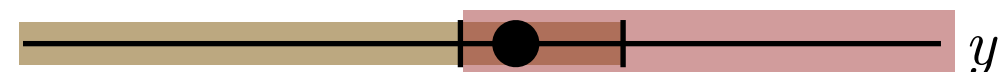
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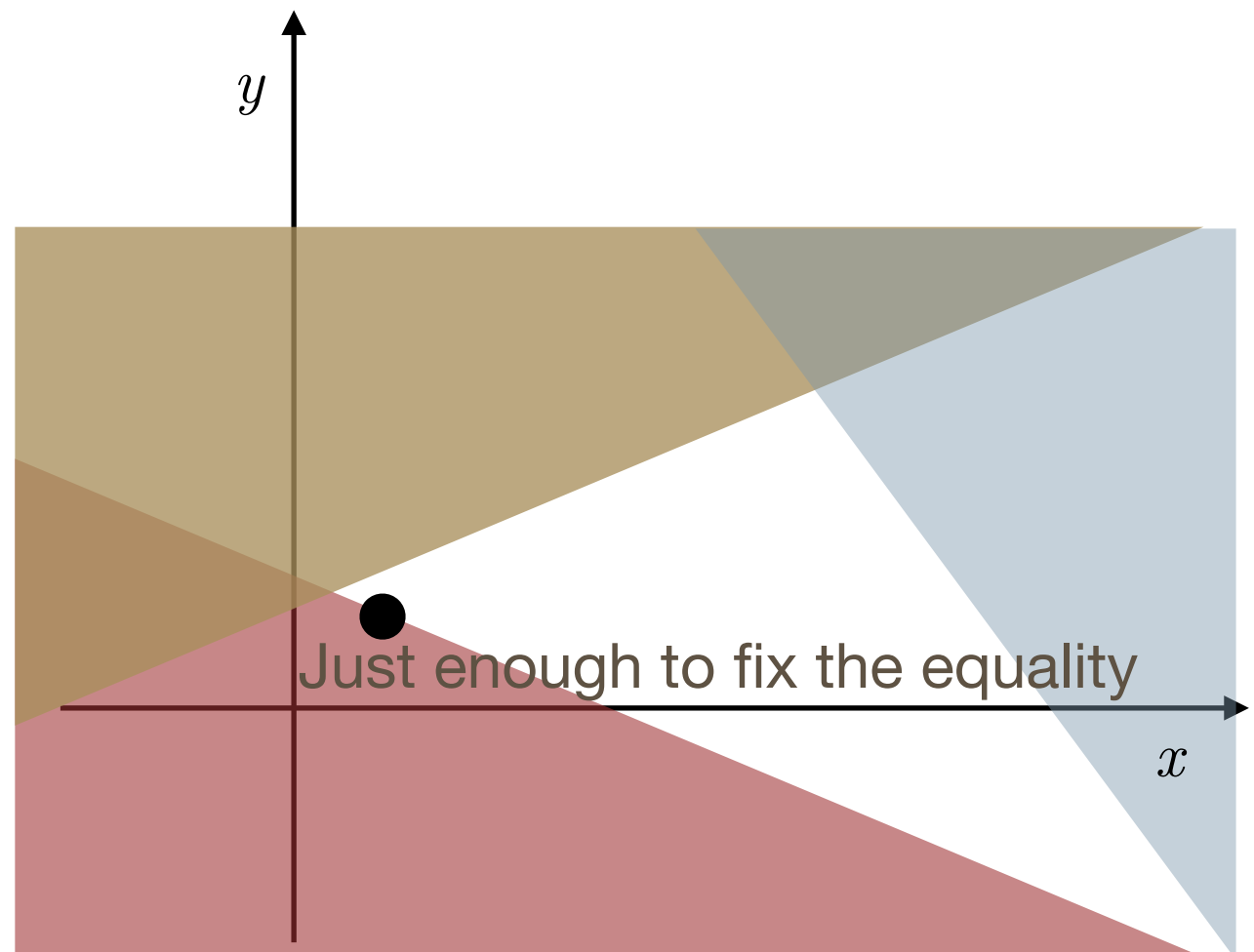
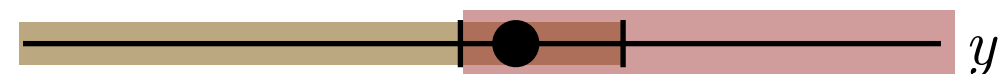
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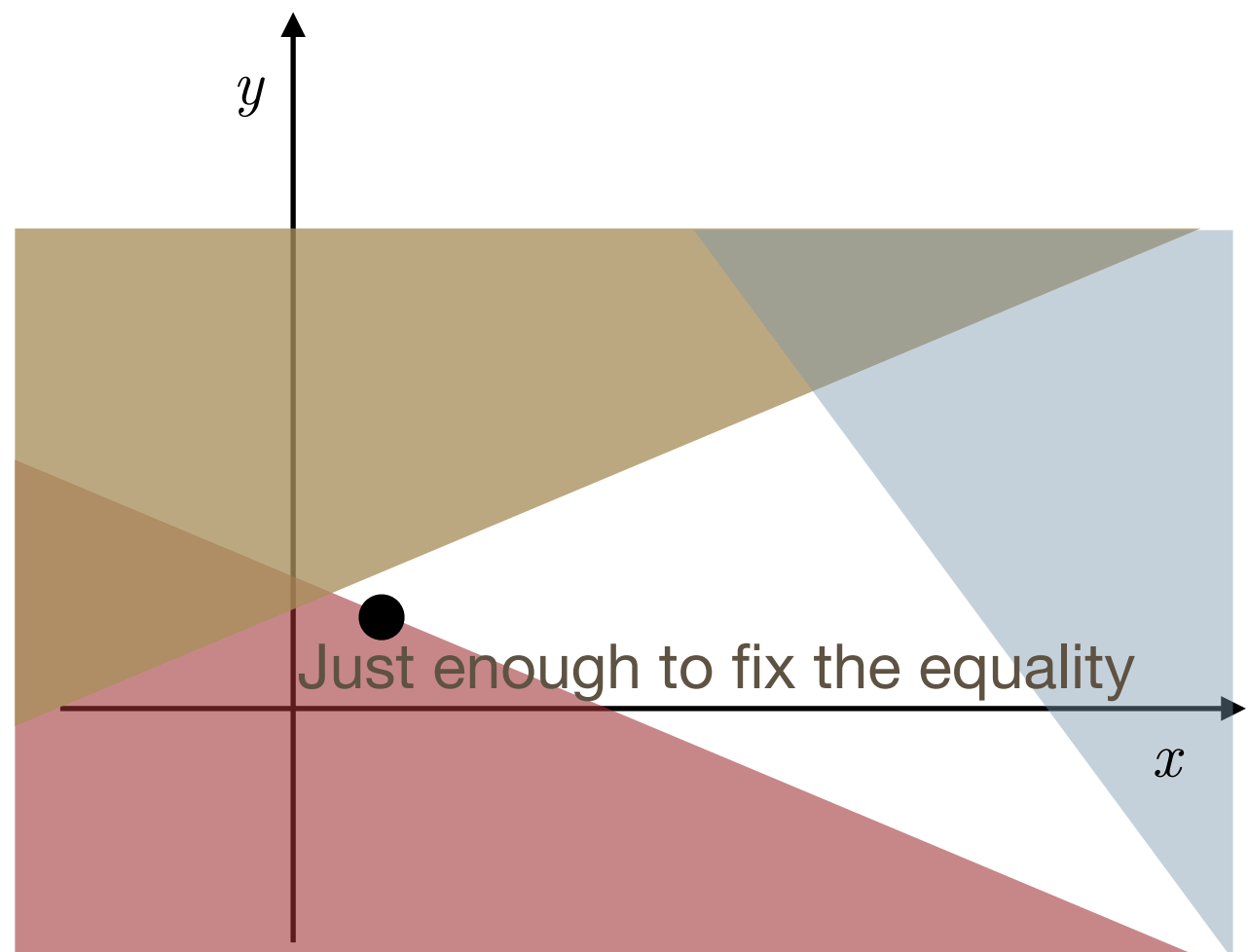
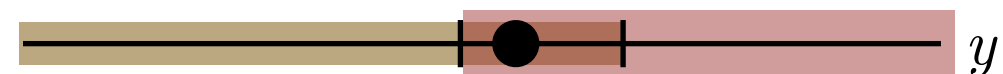


Simplex Example



Simplex Example

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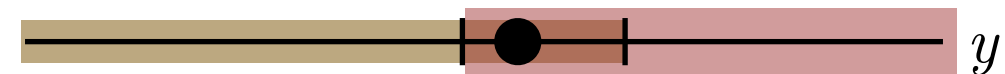


Simplex Example


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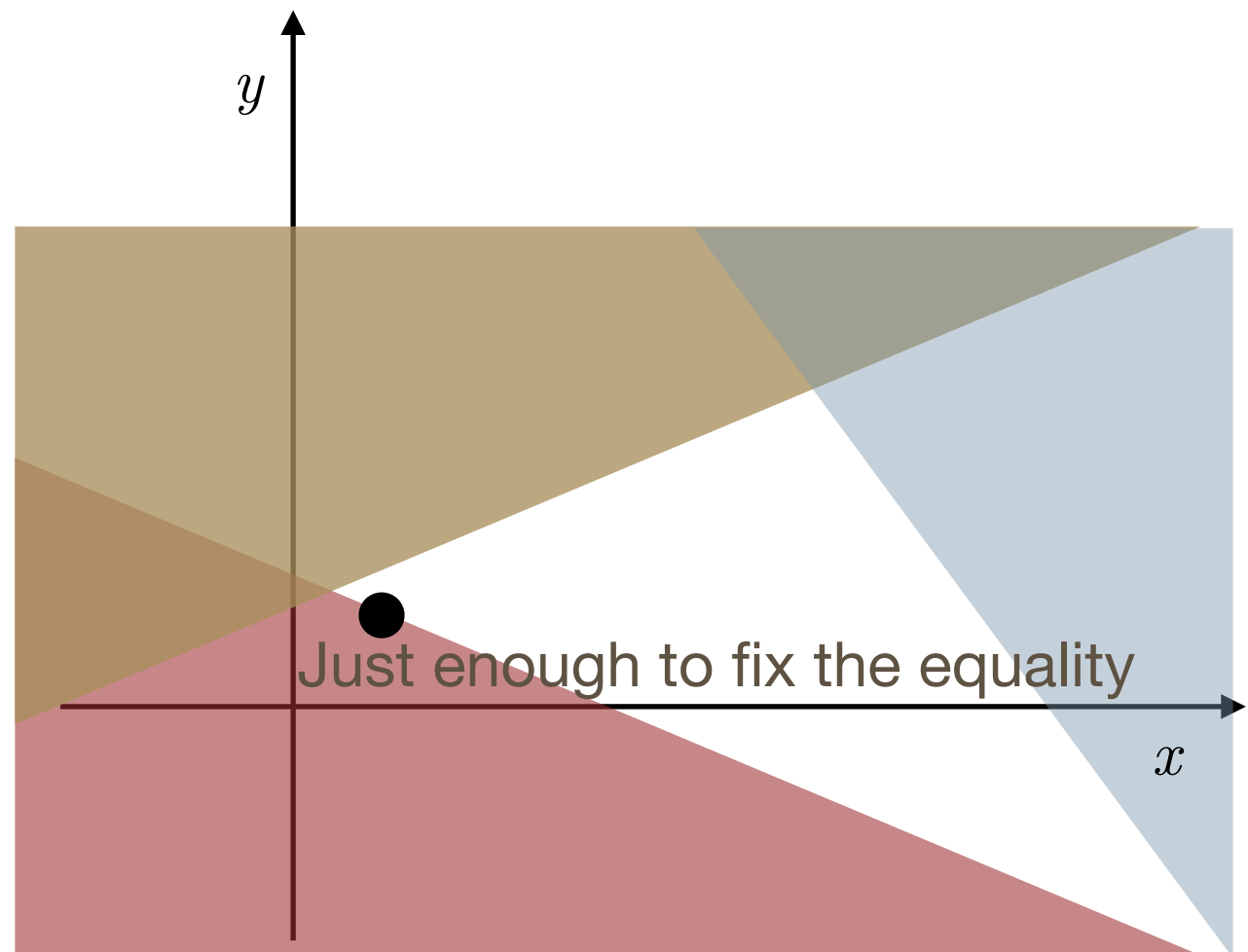
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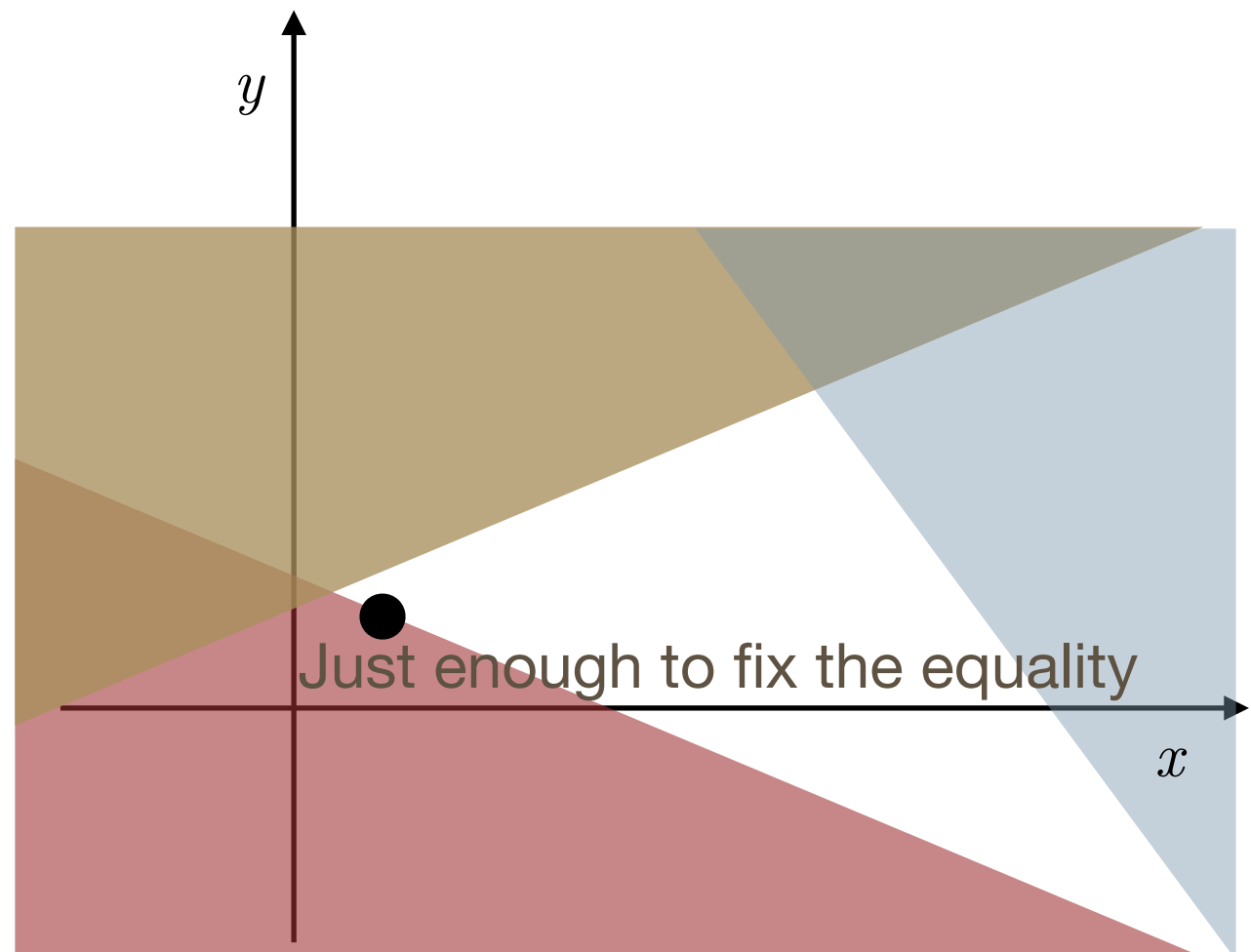
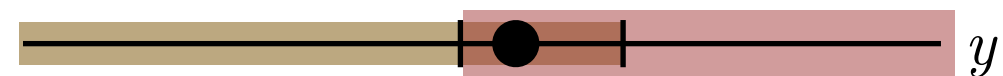
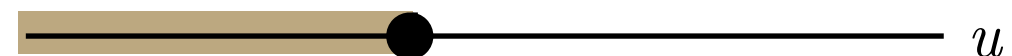
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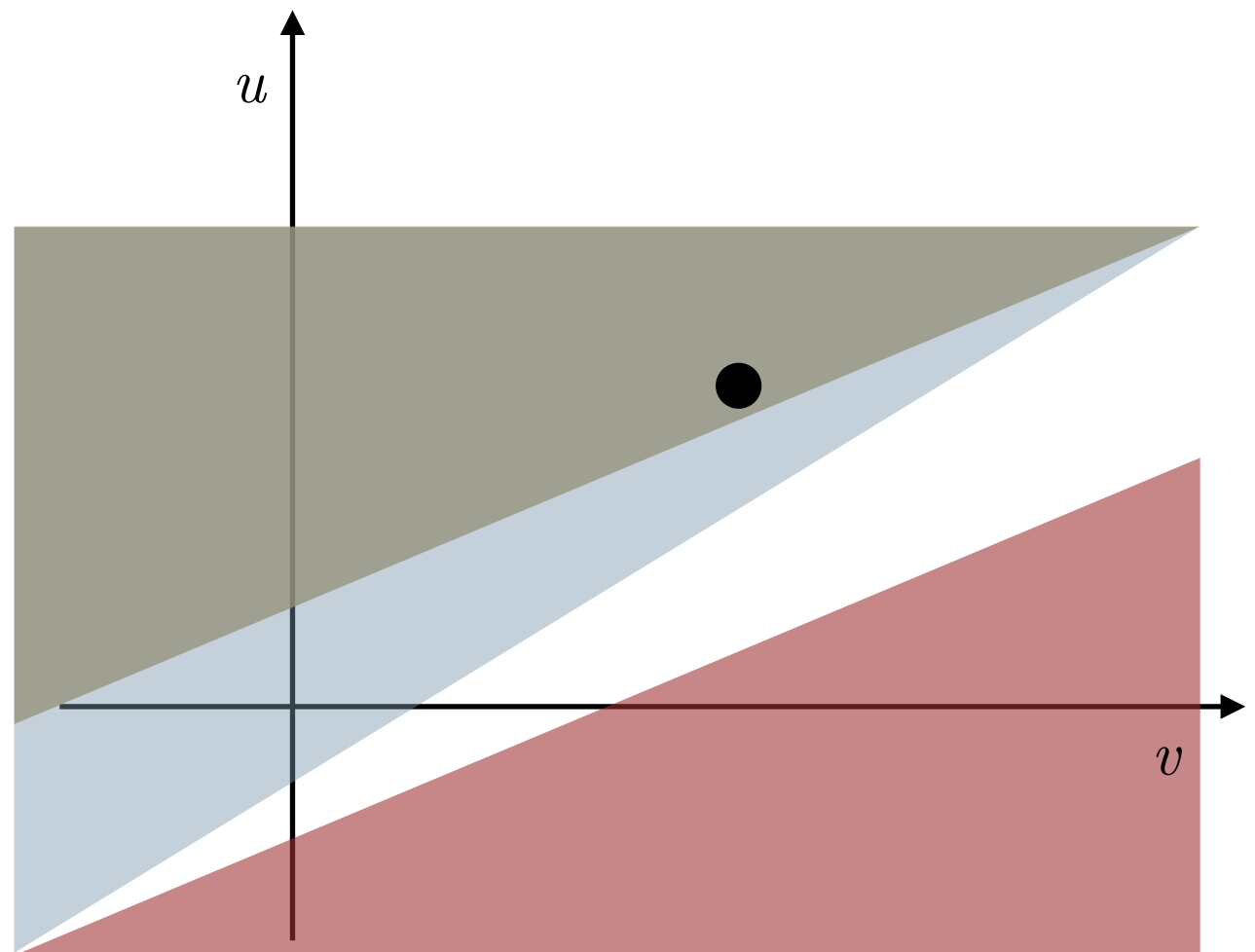
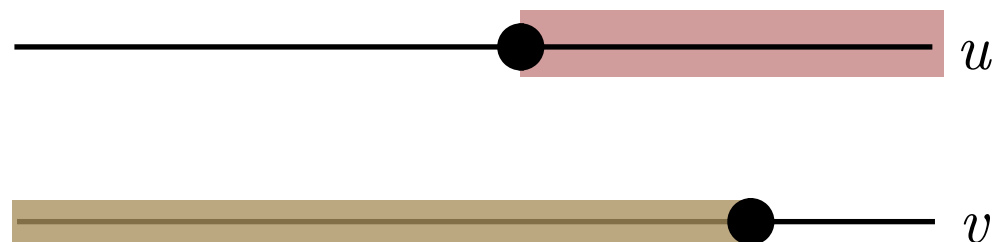
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1. The expressions change
2. One variable is fixed to its bound

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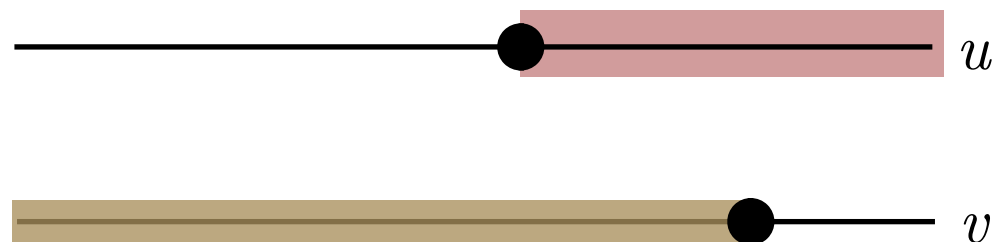
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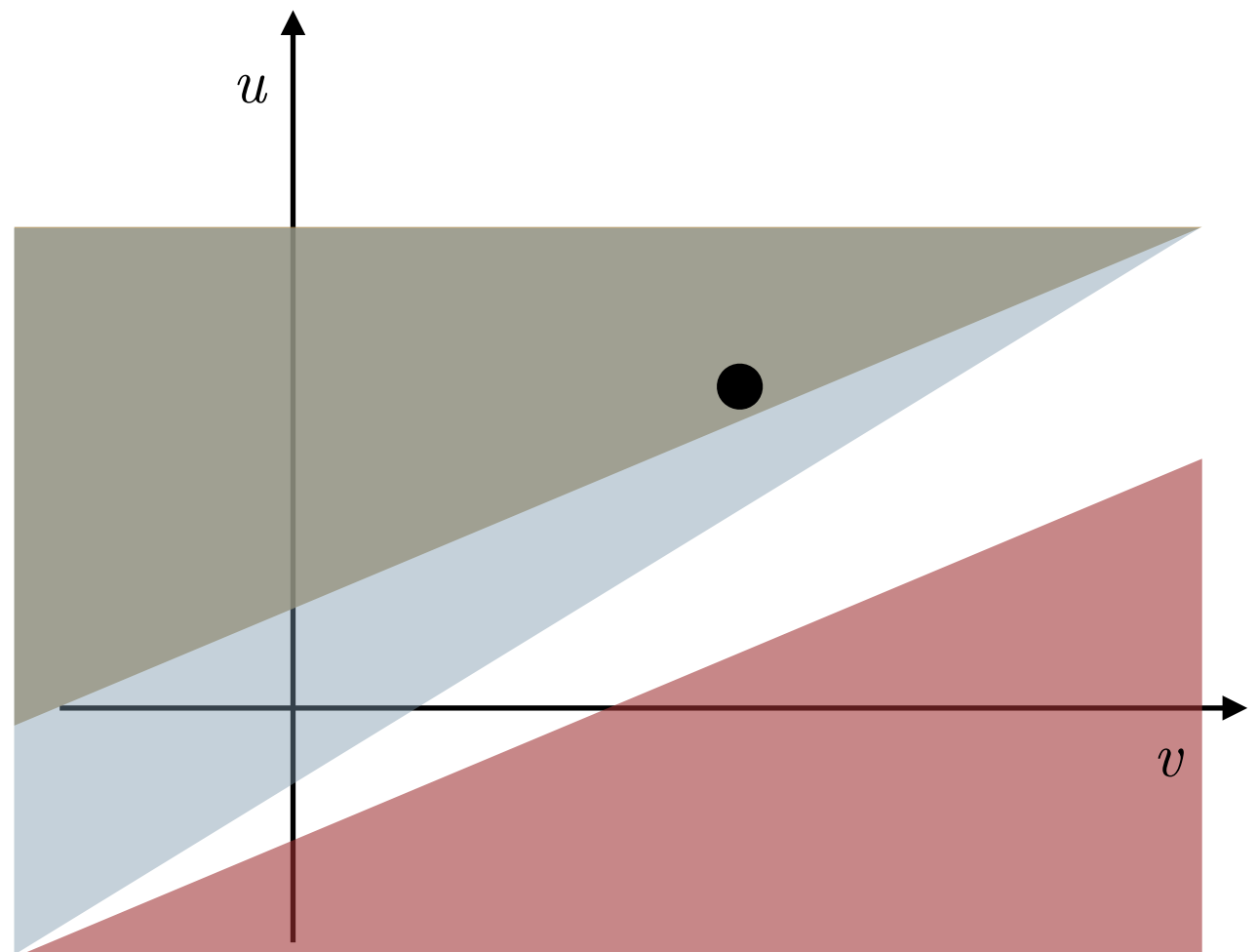
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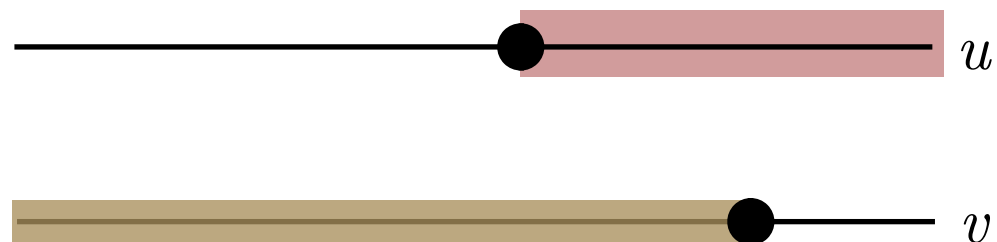
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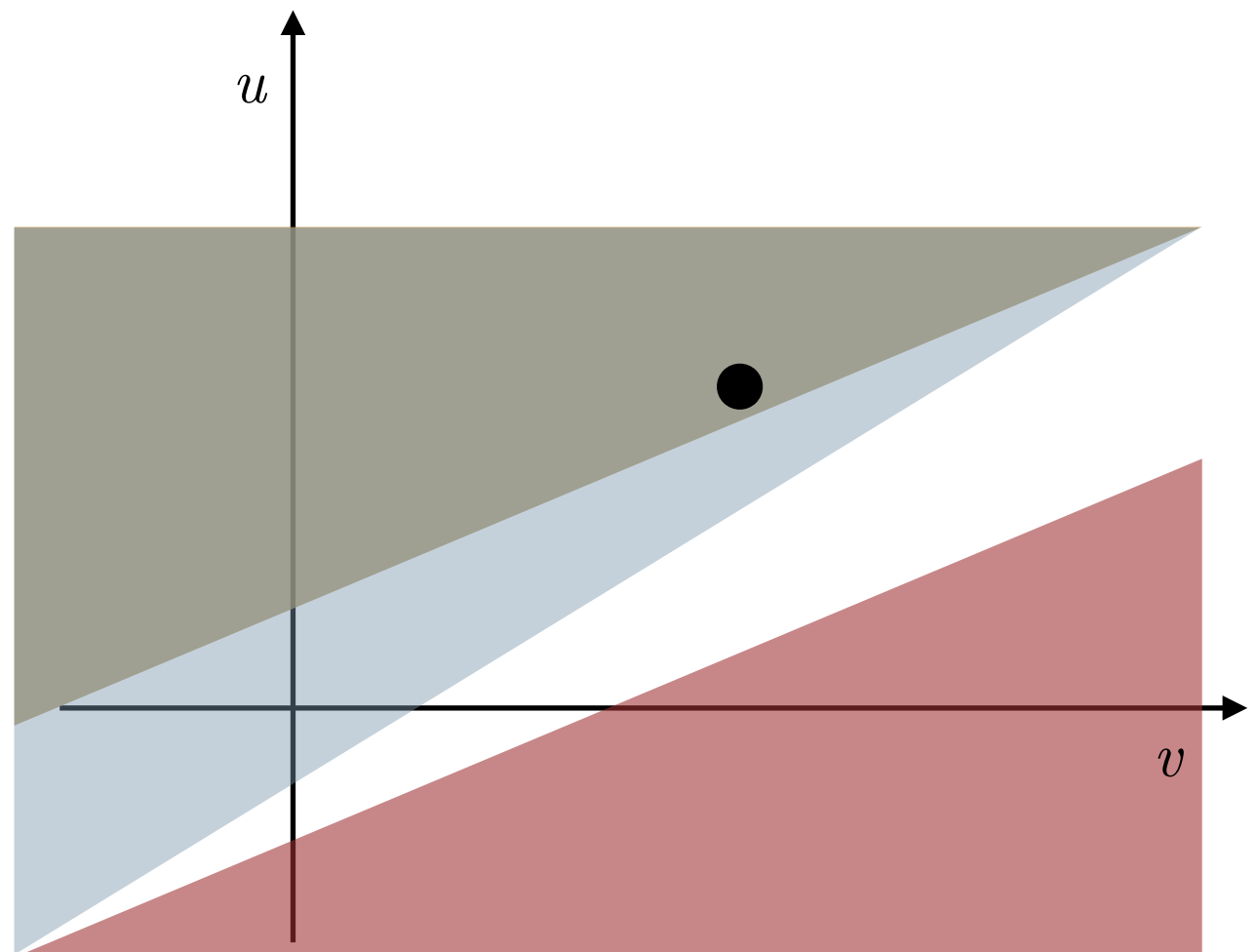
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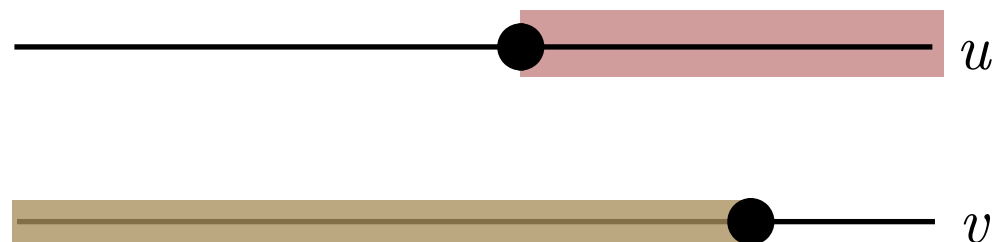


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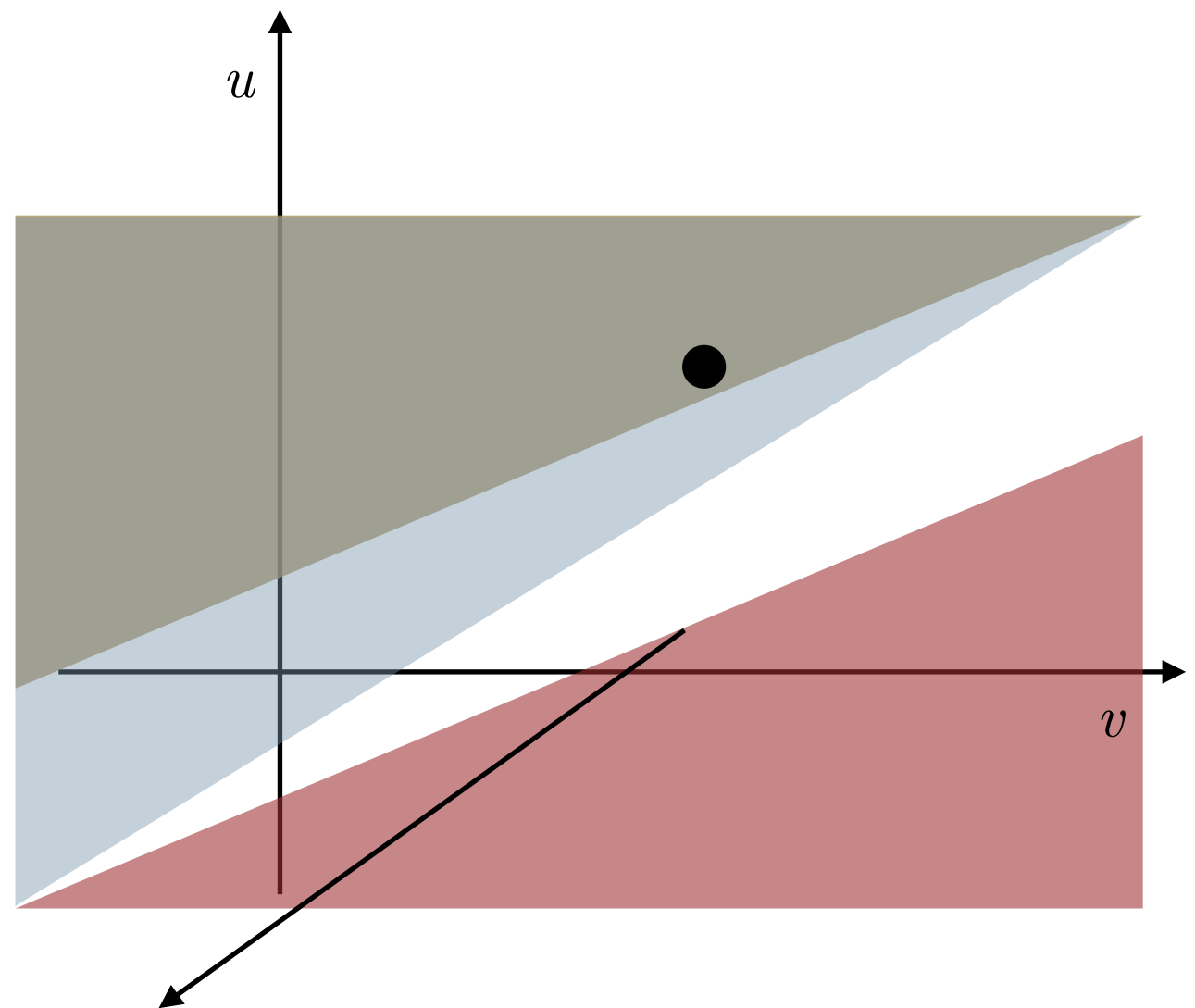
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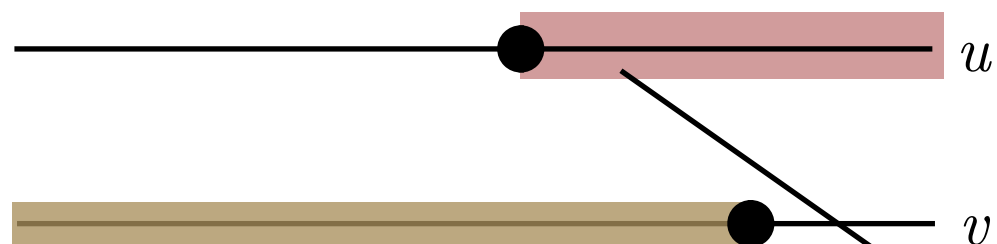
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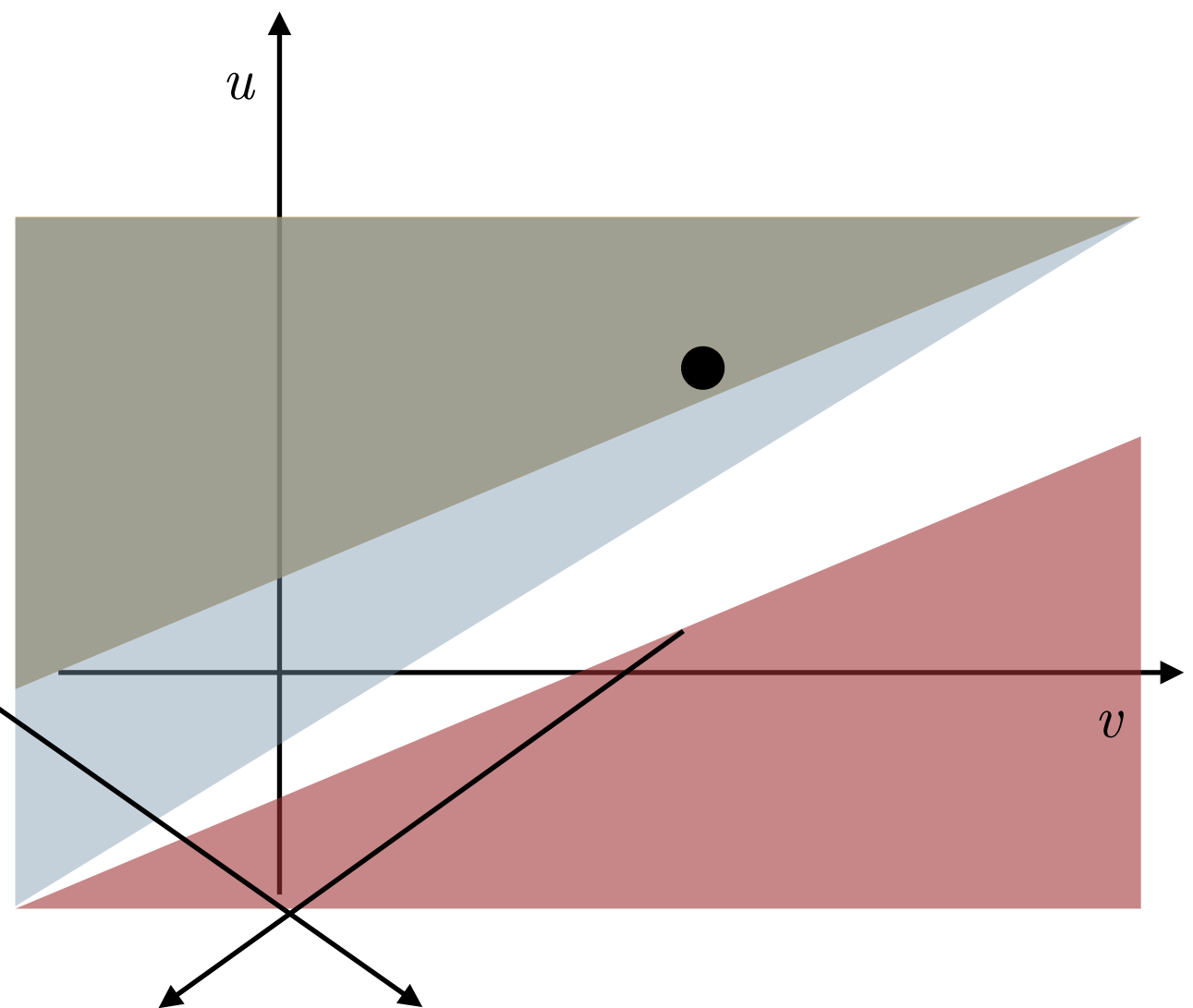
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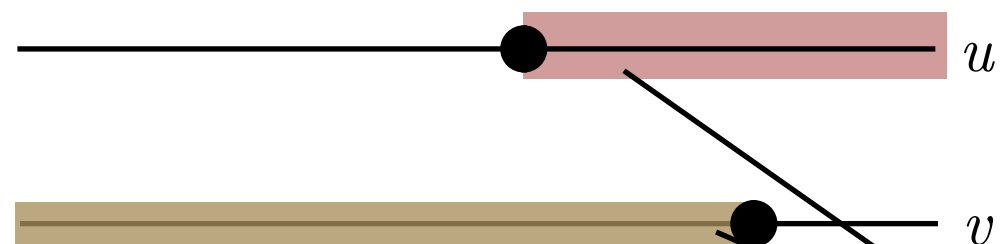
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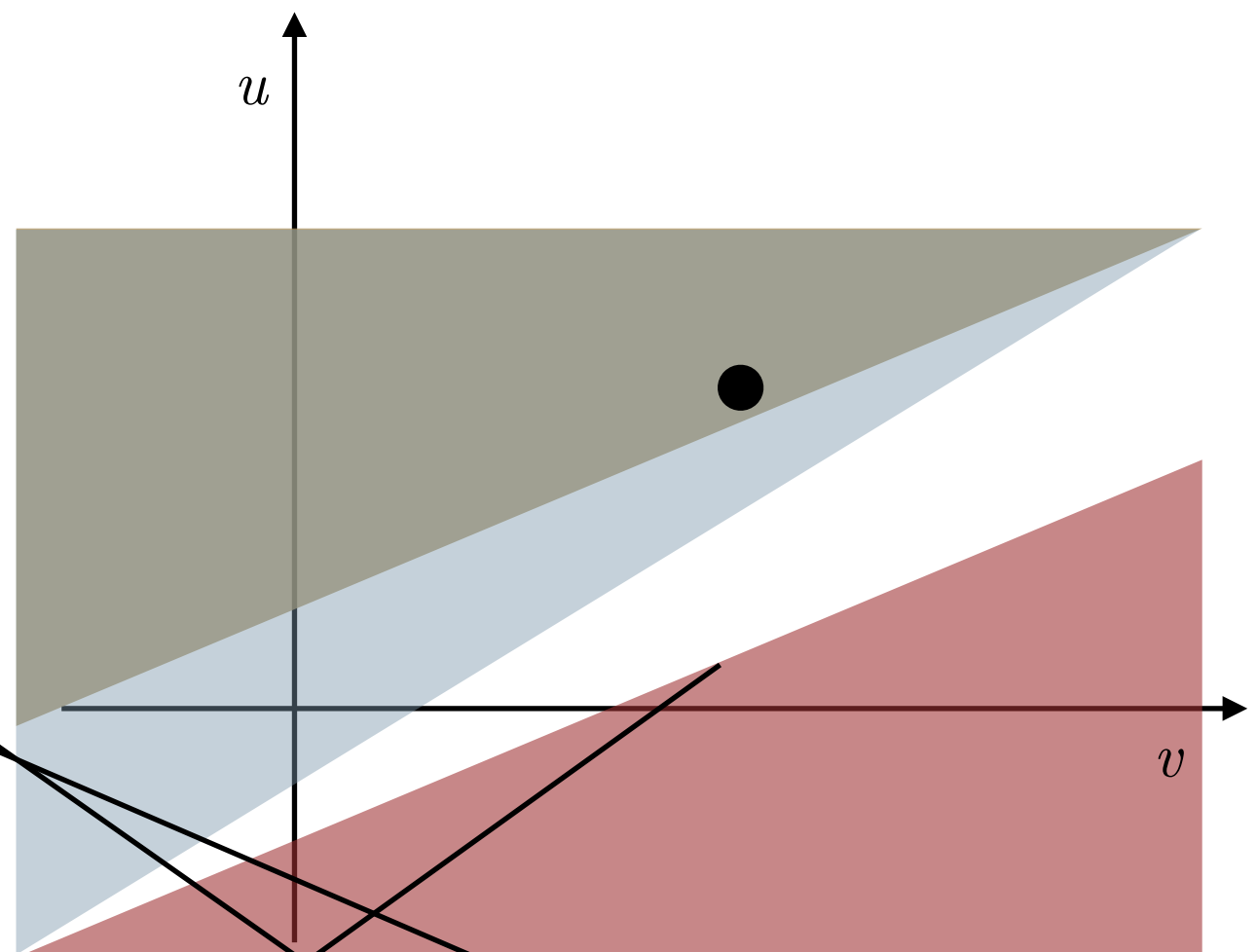
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LRA Interpolation

Assume that the expression
bound that could not be satisfied
was $1 > 0.5v - u$

and the bounds for the variables
 u, v were

$$u > 3$$

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Assume that $(u > 3) \in B$ and
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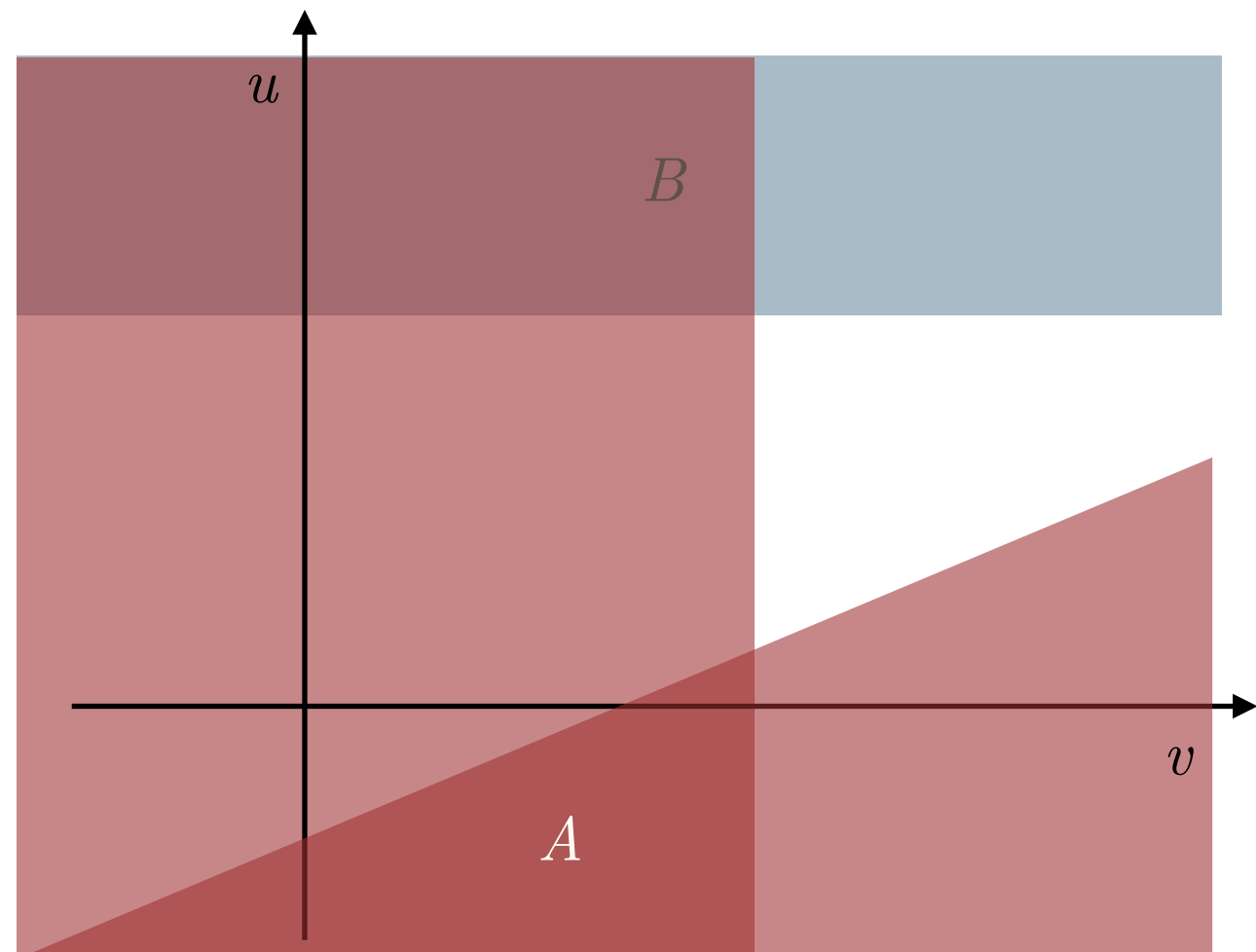
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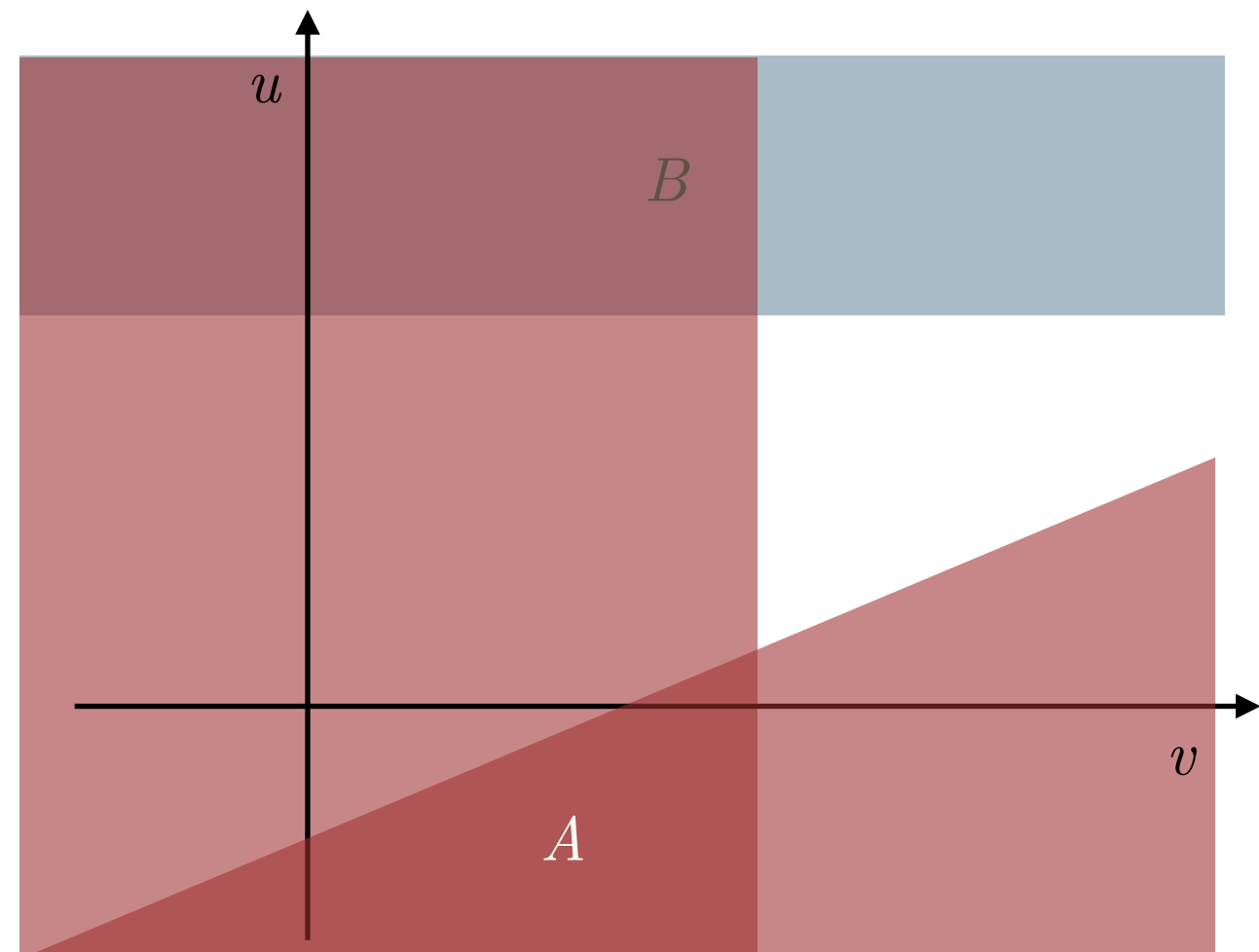
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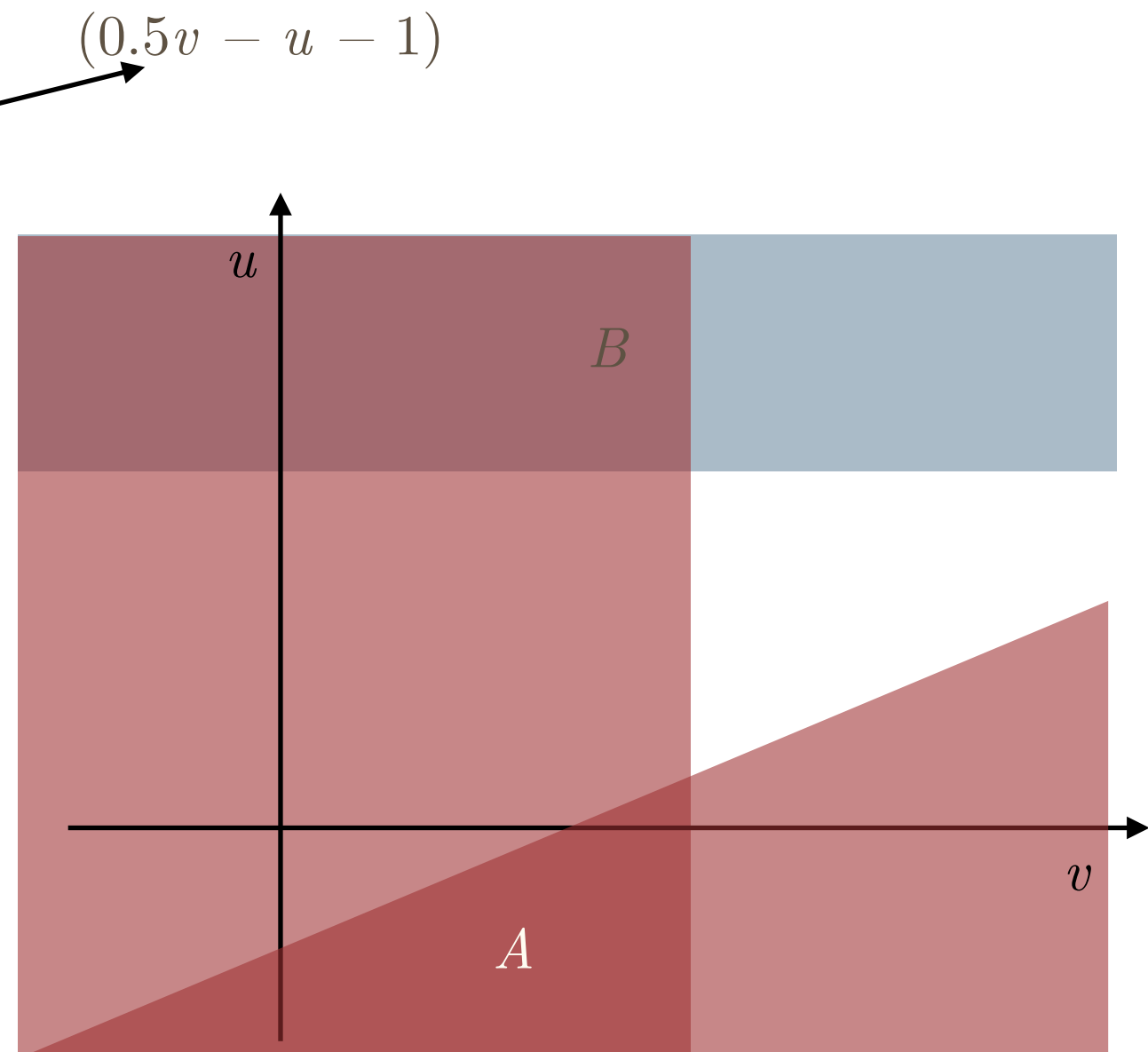
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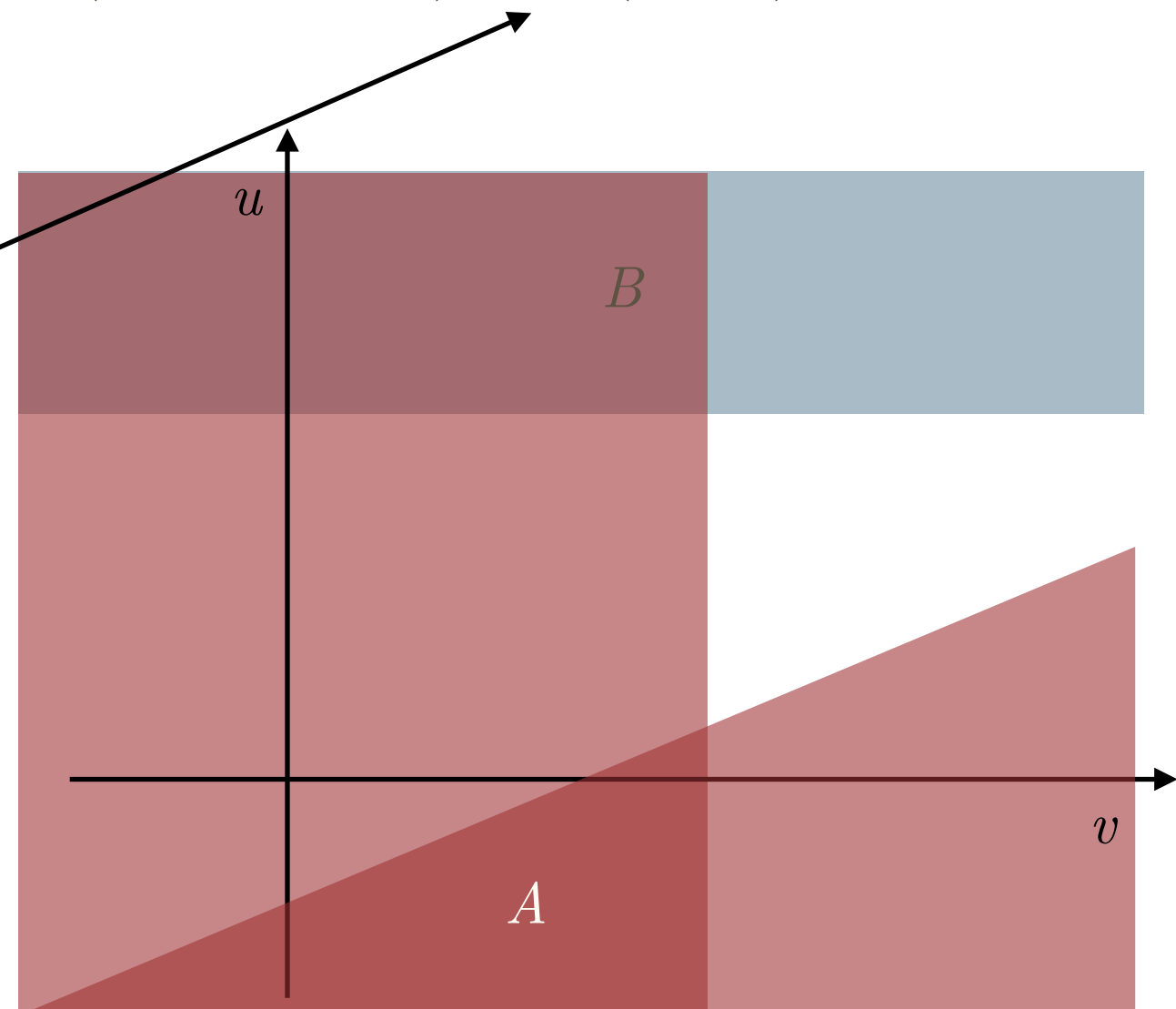
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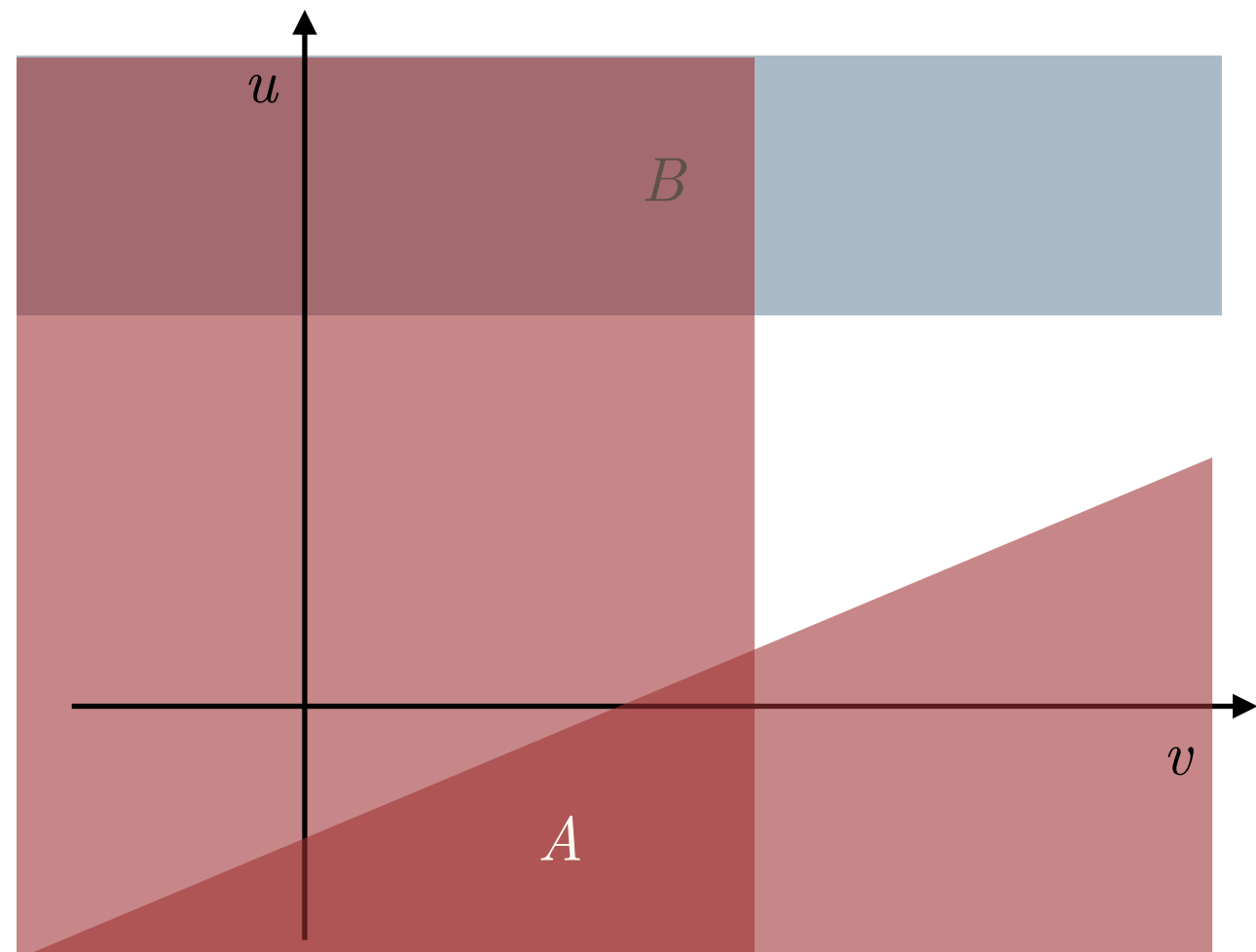
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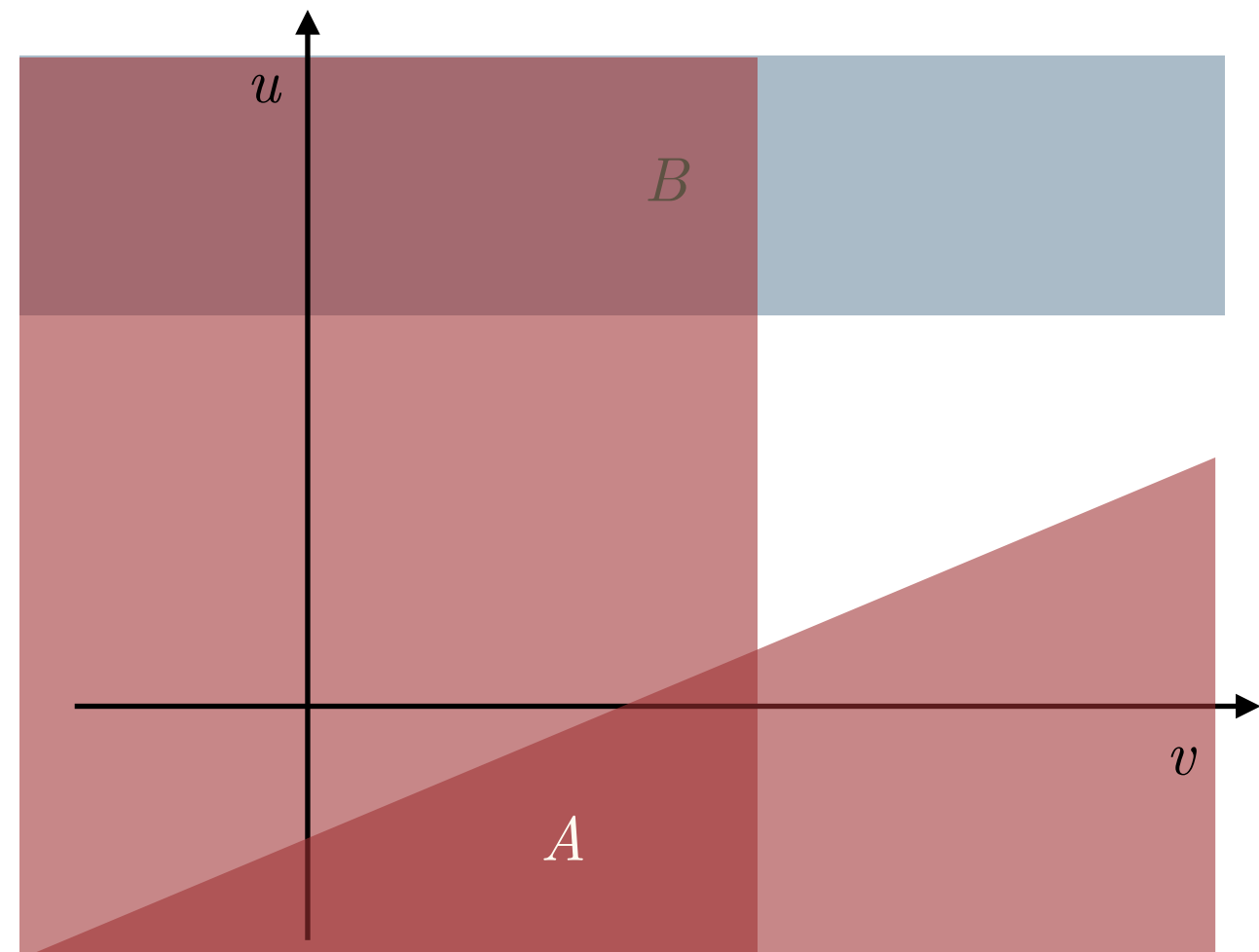
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$$-u - 1 + 1.5$$



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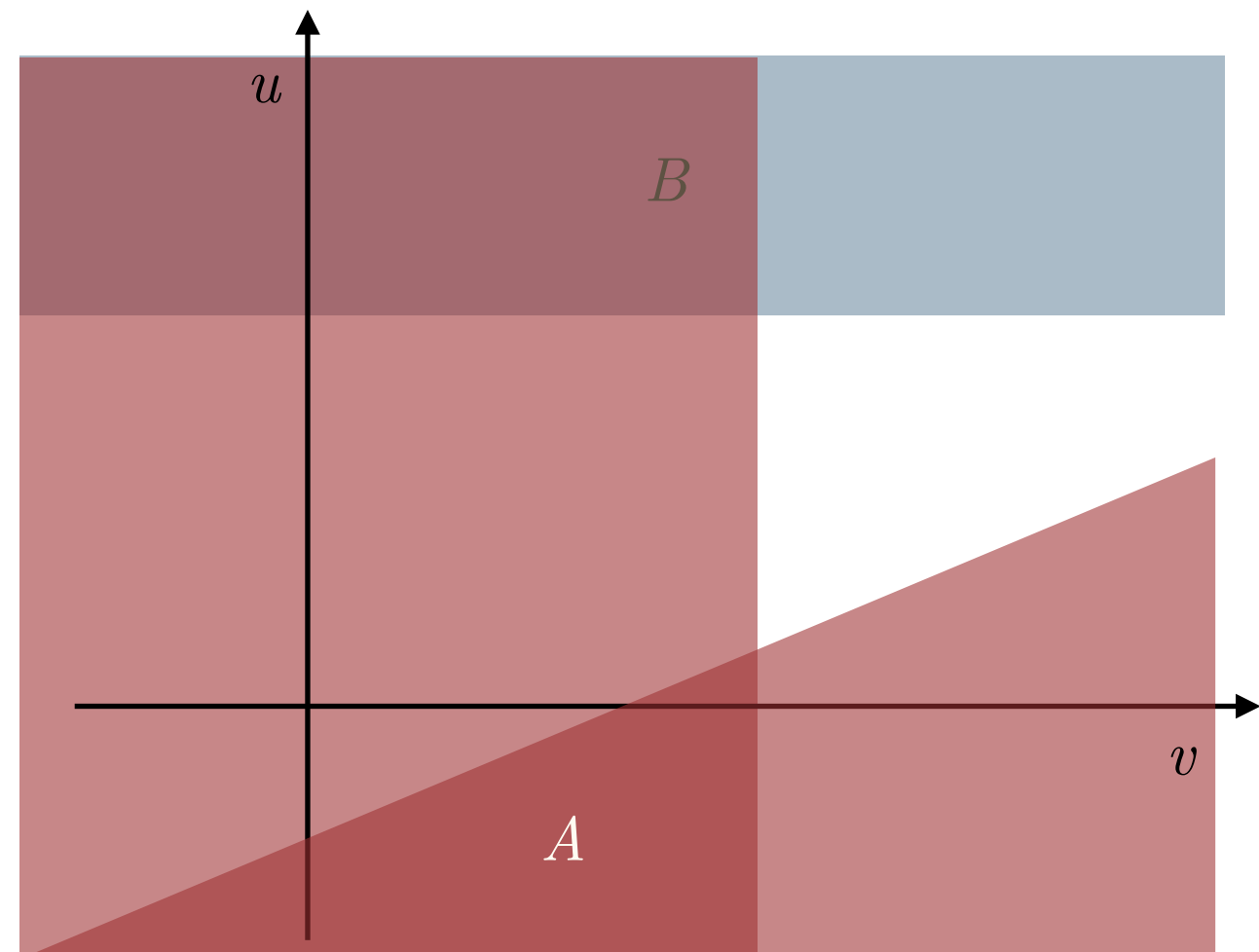
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$$v < 3$$

Assume that $(u > 3) \in B$ and
 $(v < 3) \in A$, $(1 > 0.5v - u) \in A$.

The interpolant for A is obtained by
summing to the expression the bounds in A
multiplied by their factors in the expression:

$$0 < -u - 1 + 1.5$$



LRA Interpolation

Assume that the expression
bound that could not be satisfied
was $1 > 0.5v - u$
and the bounds for the variables
 u, v were

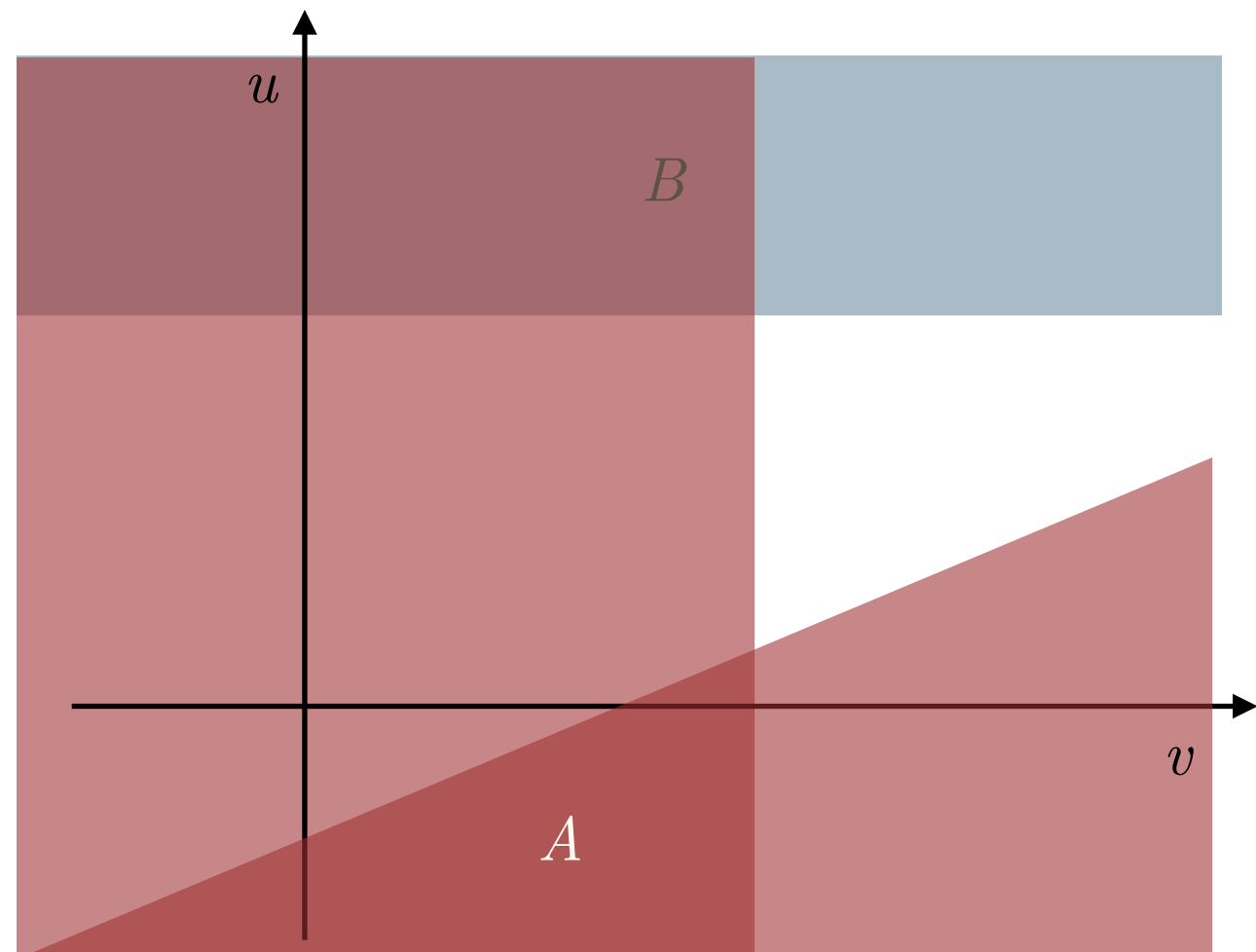
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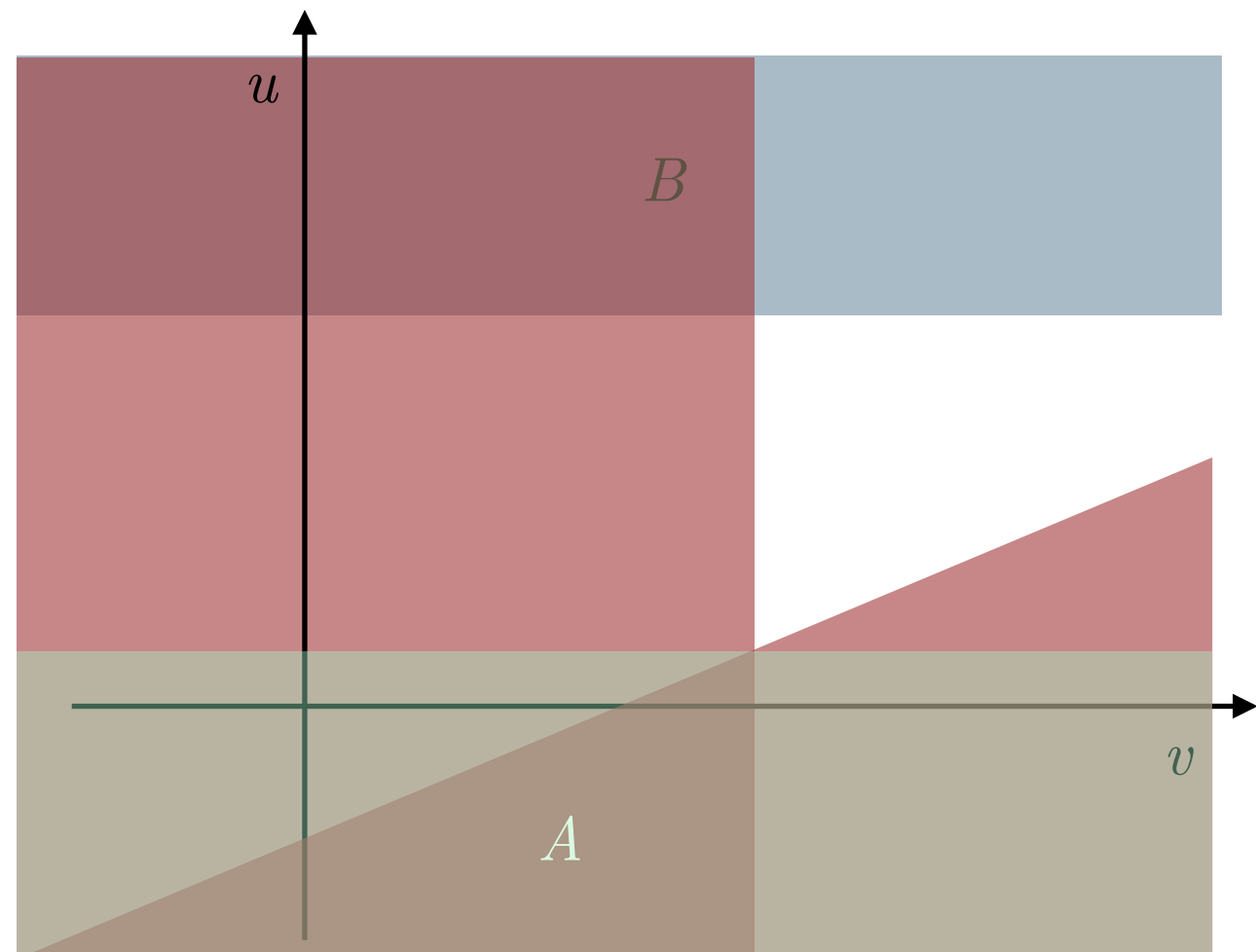
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Duality-based Interpolation for LRA

Given a primal interpolant

$$I = c_1 \leq t(\boldsymbol{x}),$$

the dual interpolant has the form

$$I' = c_2 < t(\boldsymbol{x})$$

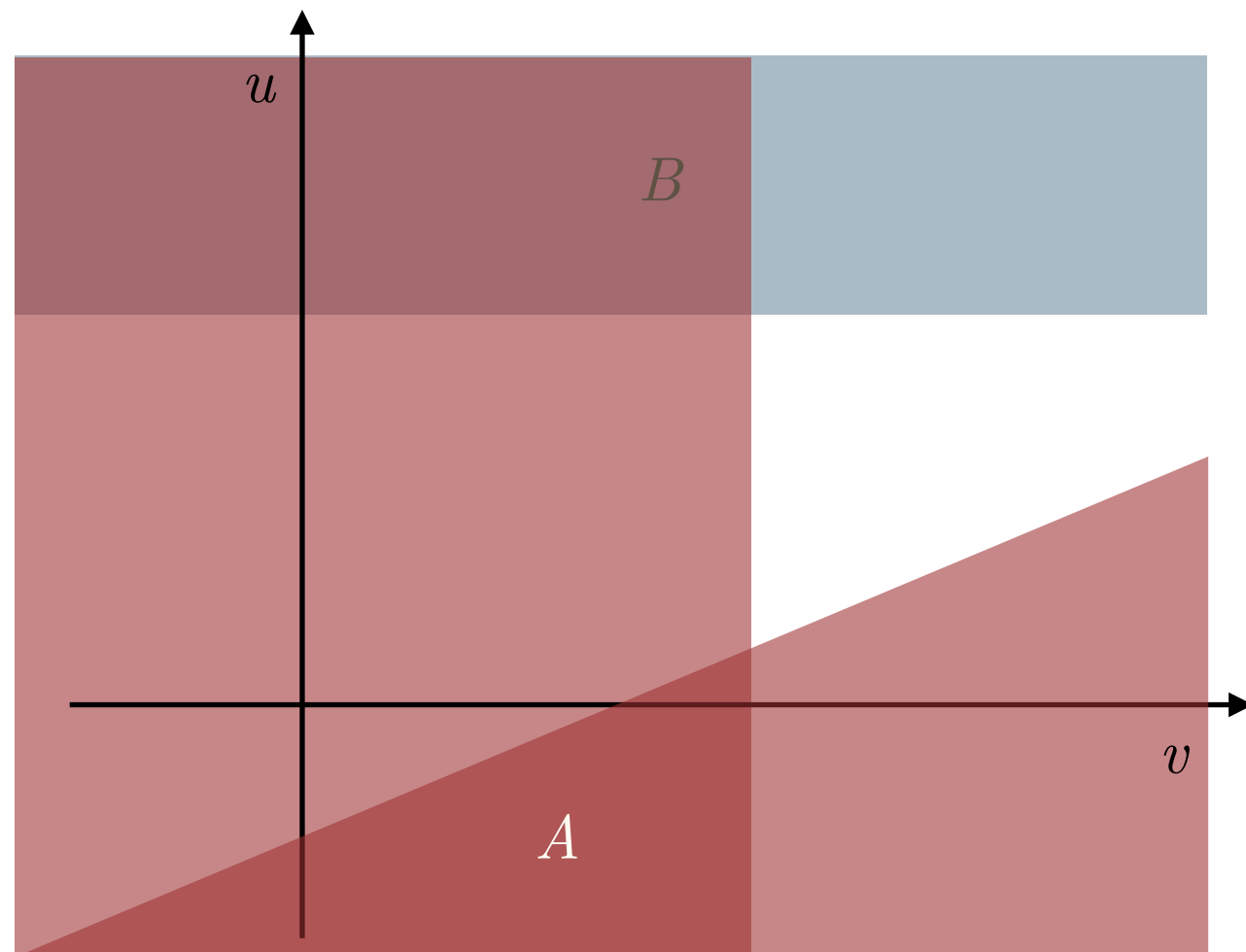
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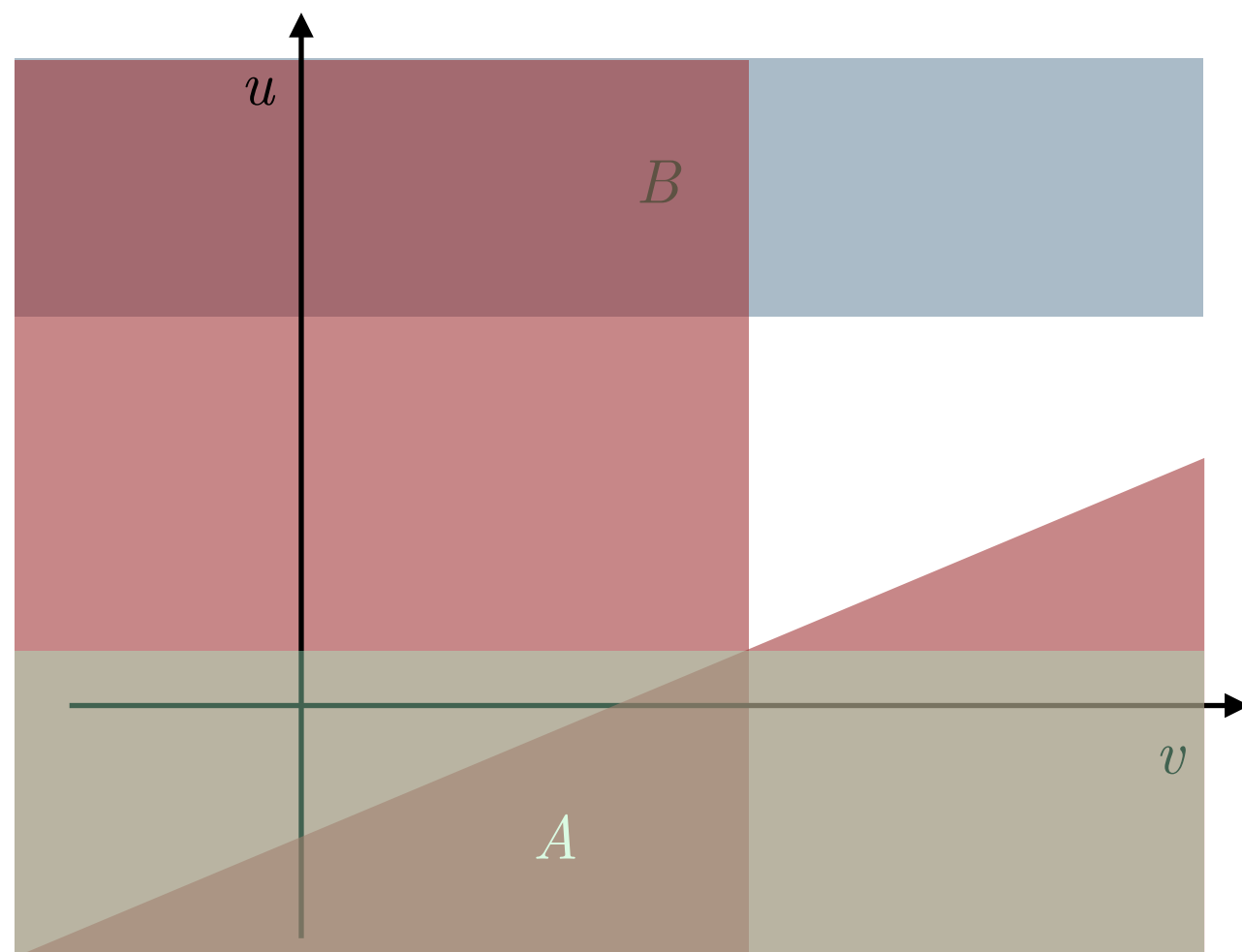
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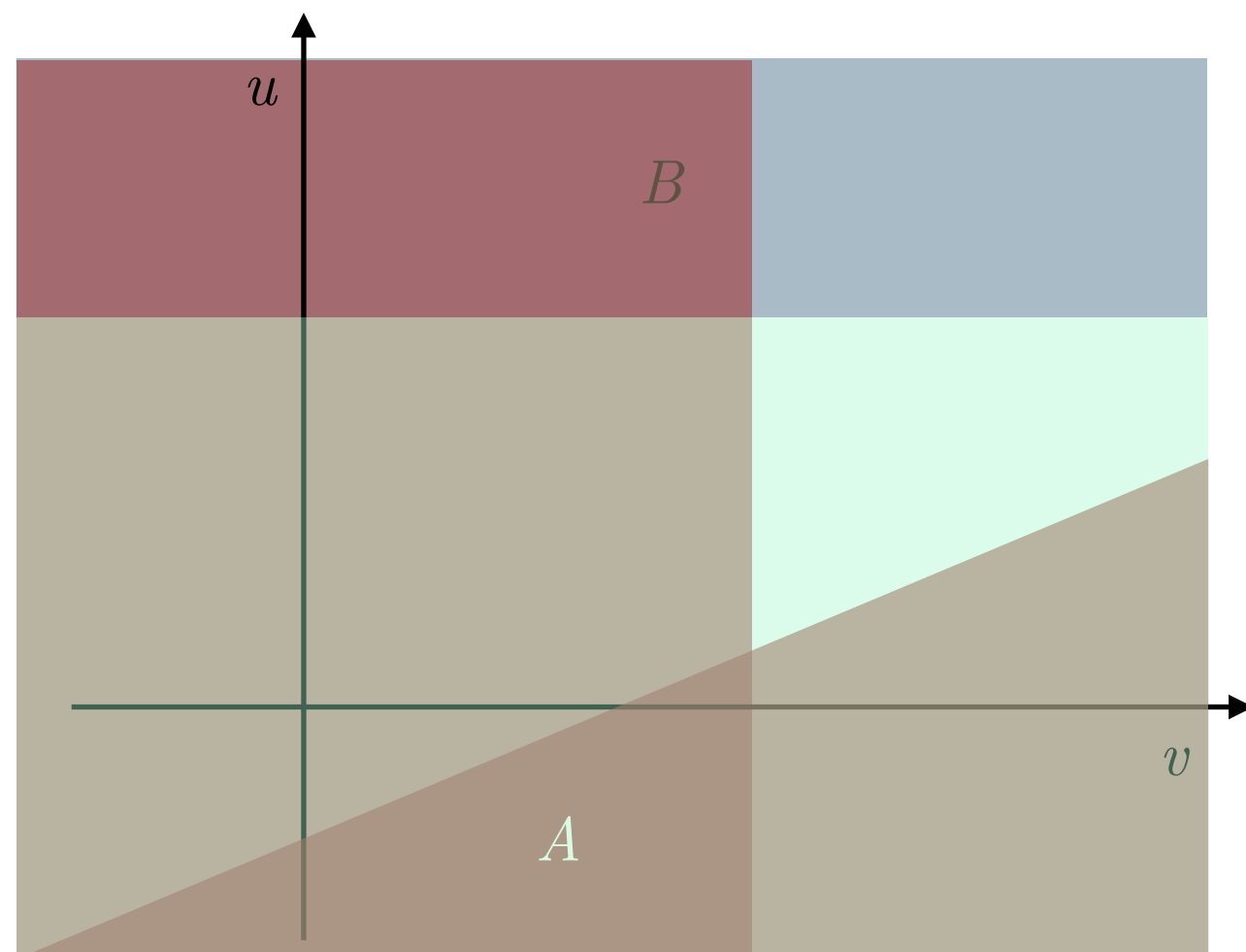
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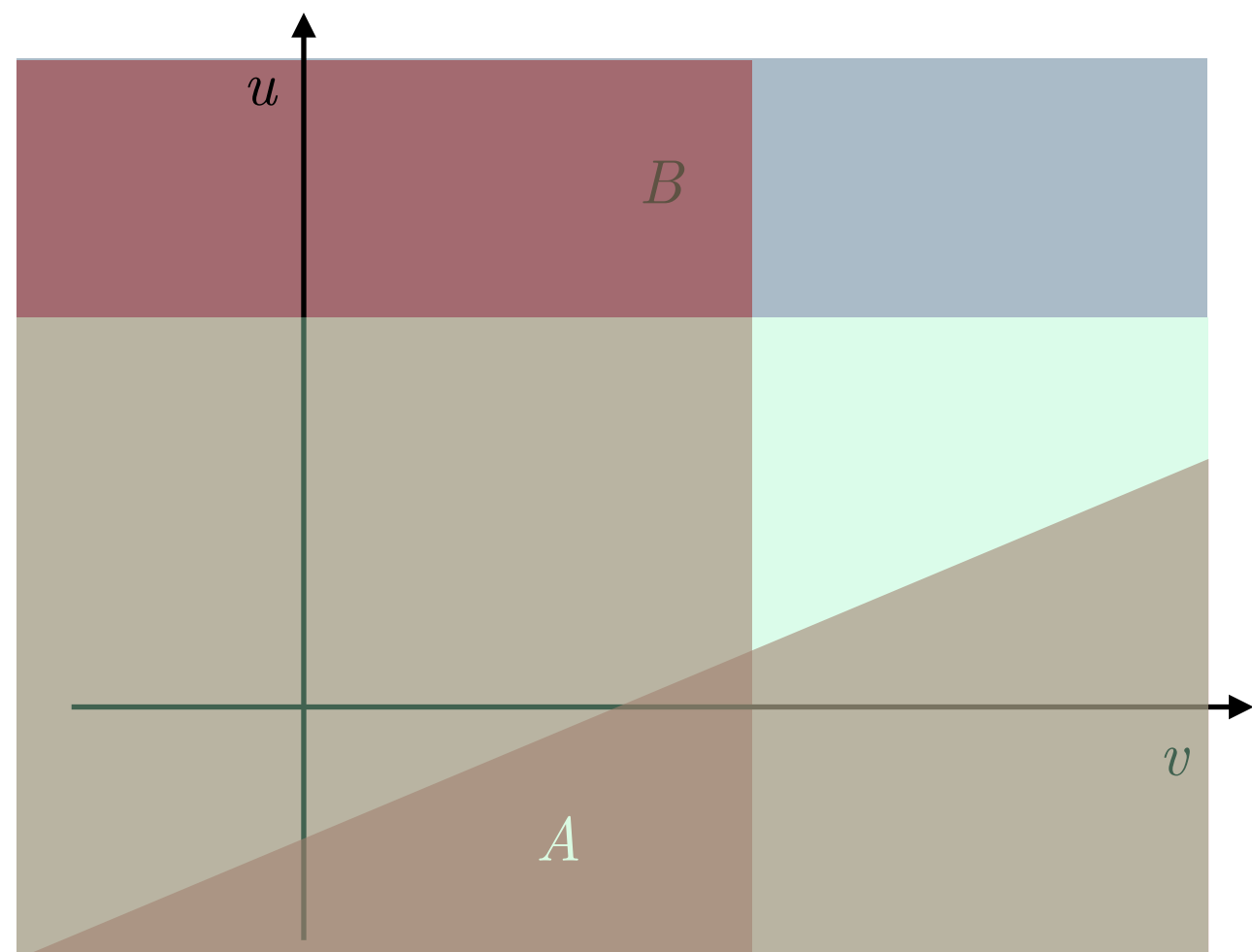
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All the inequalities of the form

$$c < t(\mathbf{x}), \quad c_1 \leq c < c_2$$

are also interpolants for A



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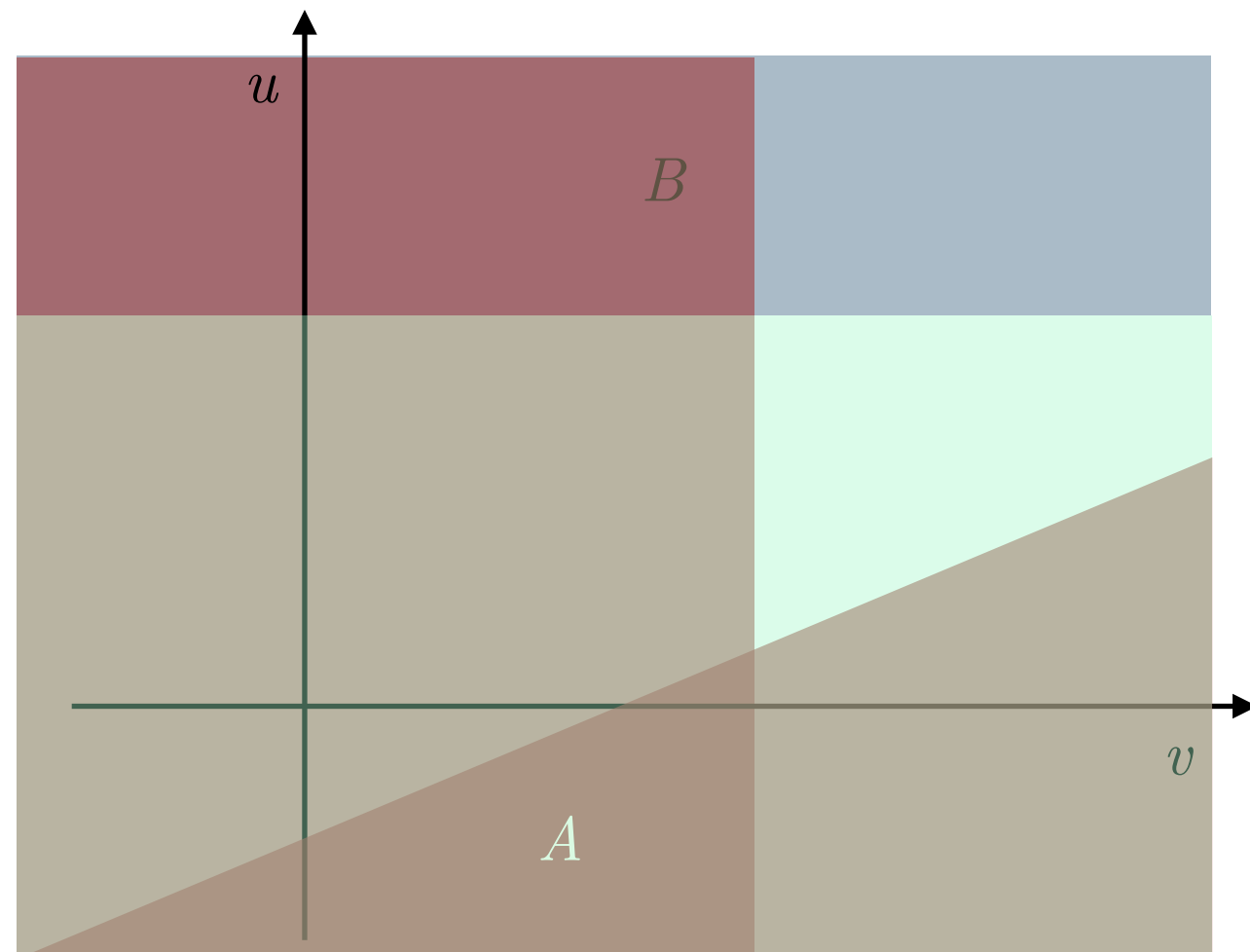
$$I' = c_2 < t(\mathbf{x})$$

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$$I \longrightarrow I'$$



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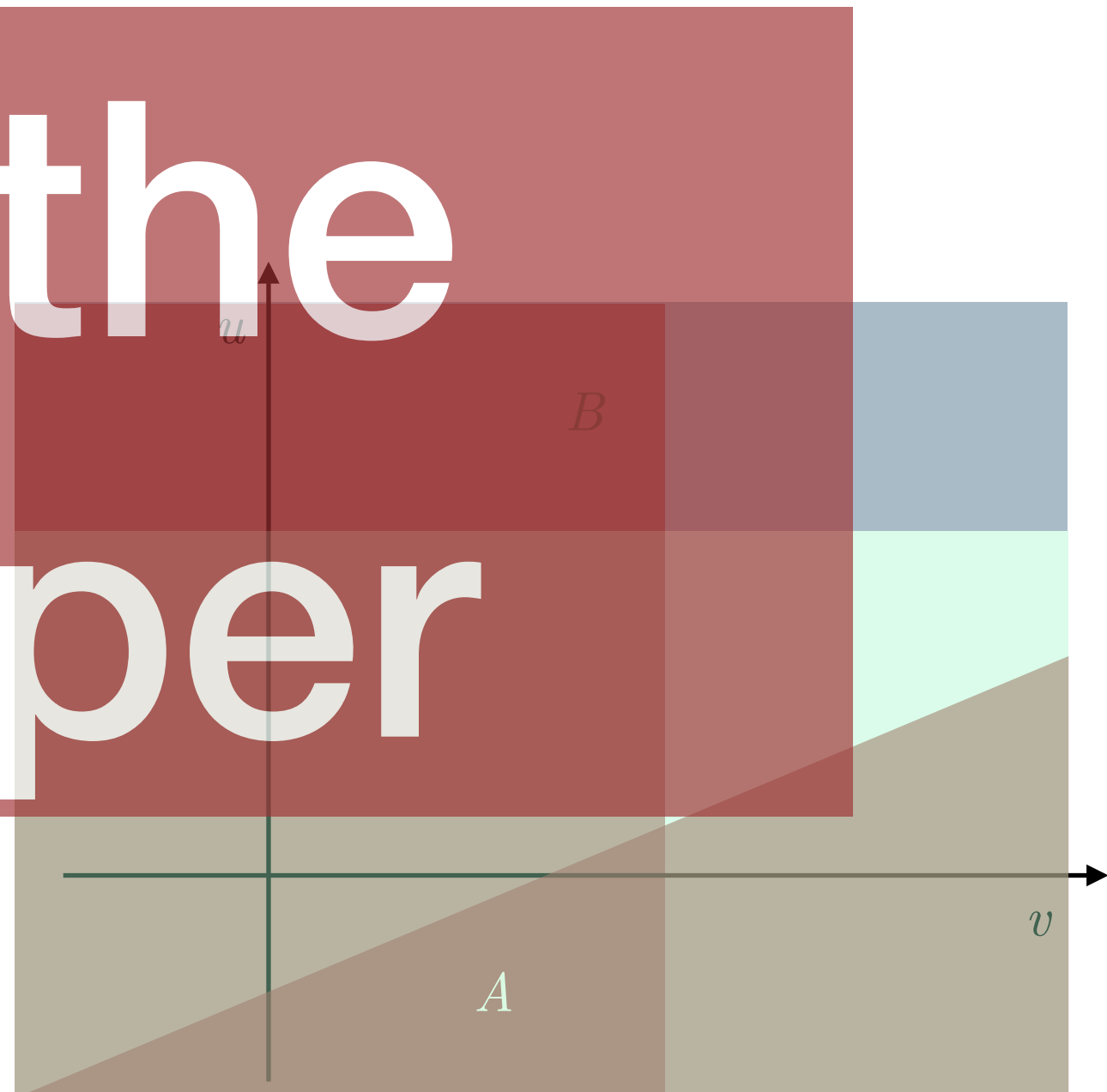
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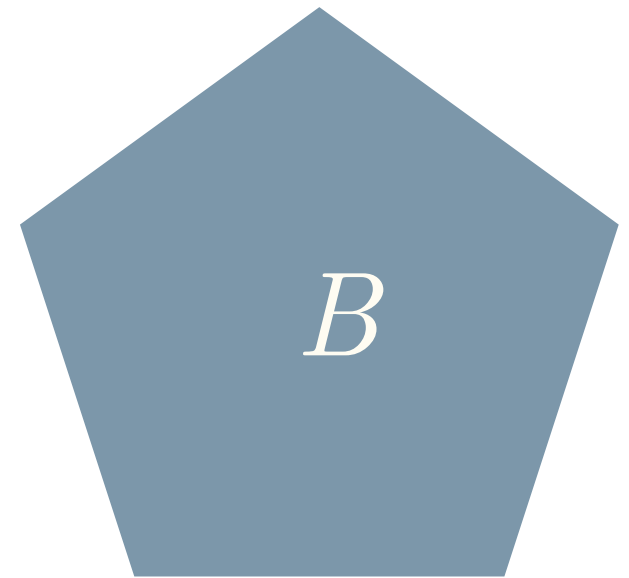
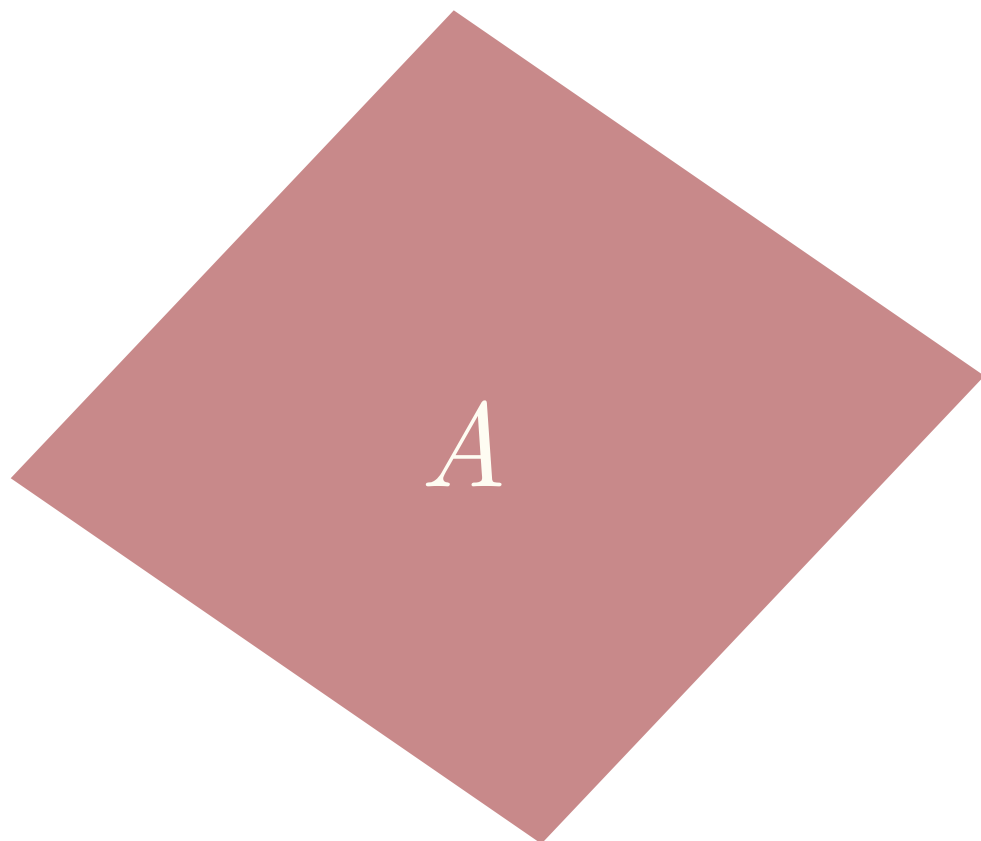
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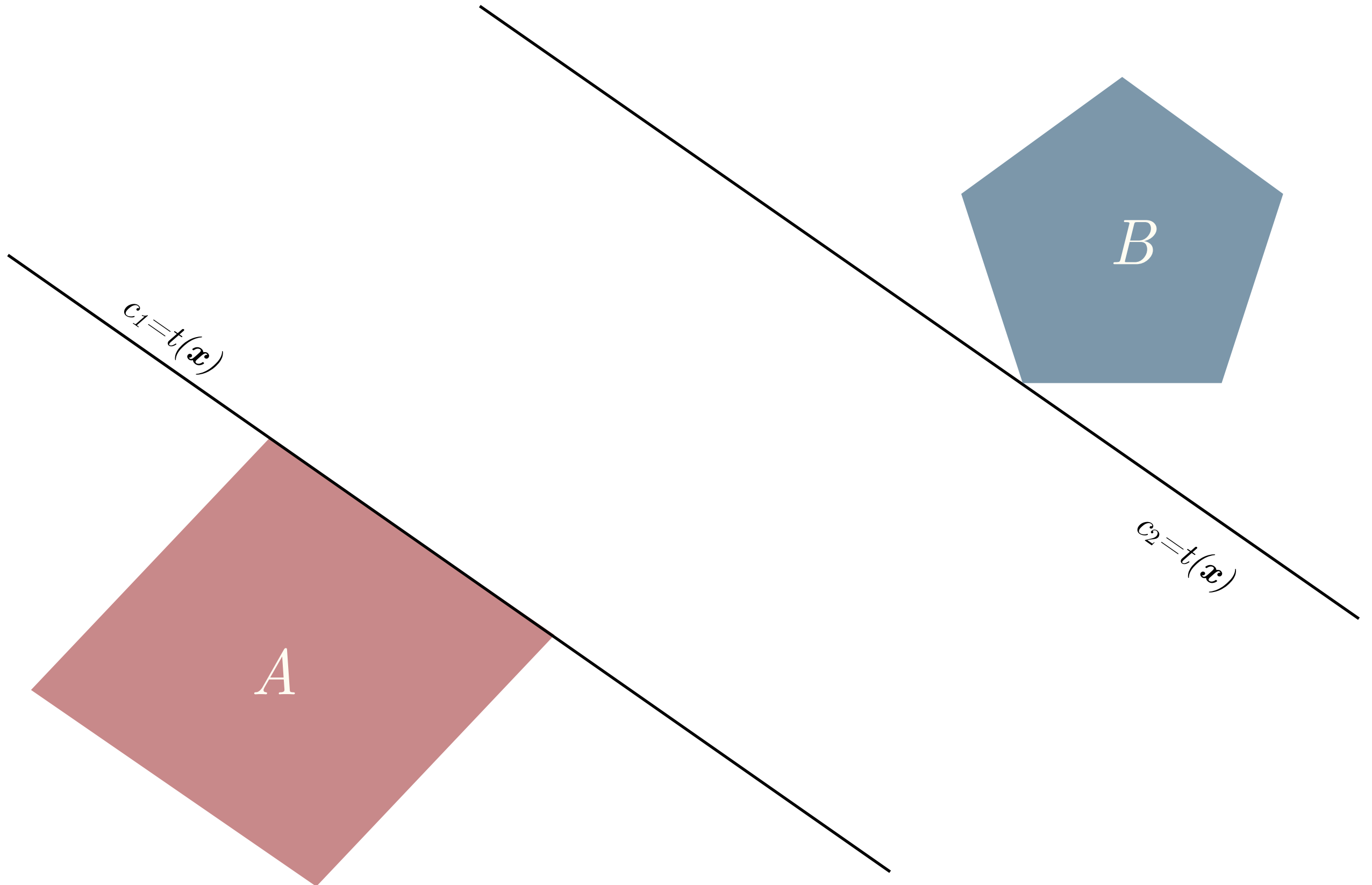
In the
paper



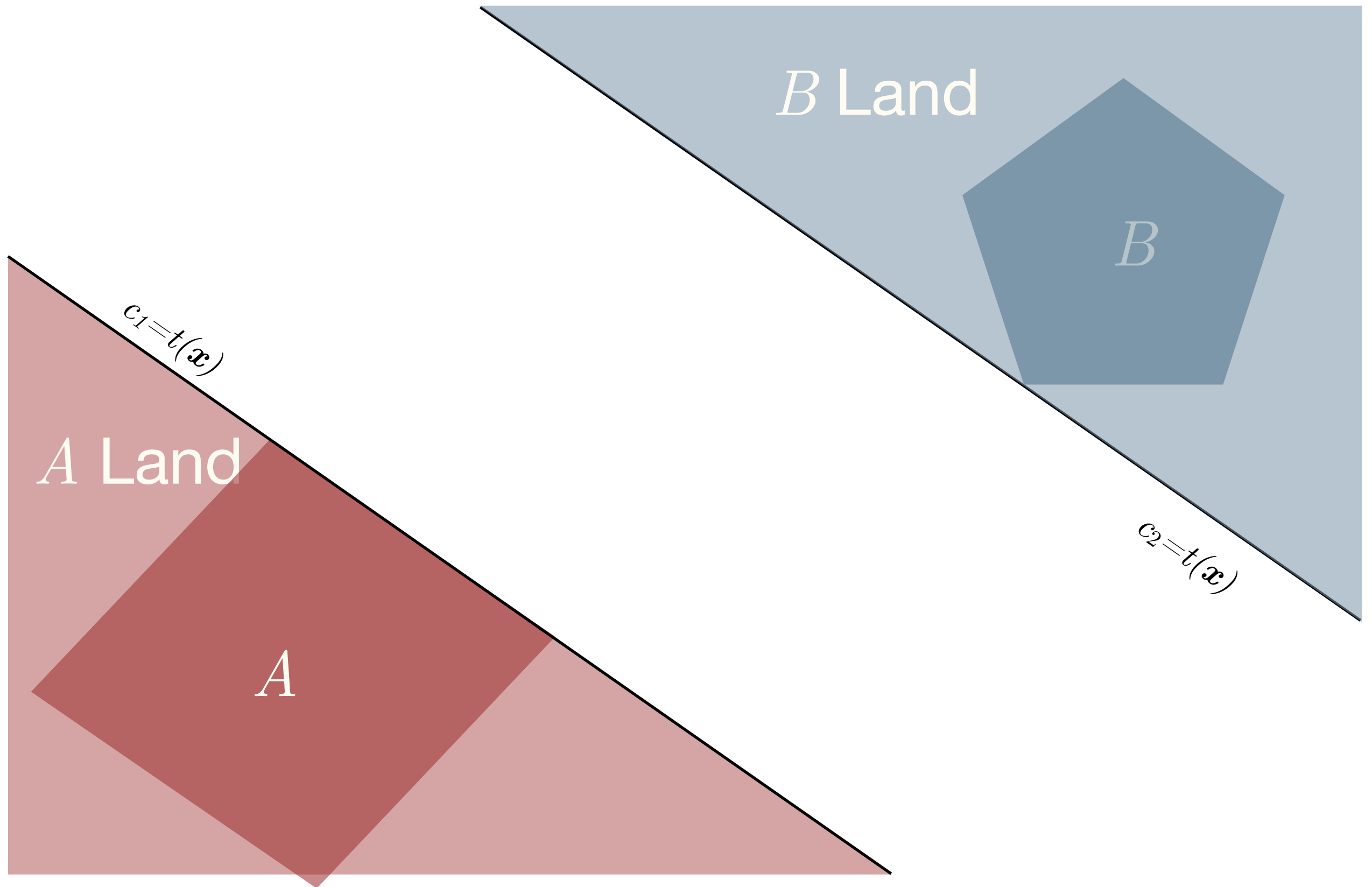
Interpolant Duality Visualized



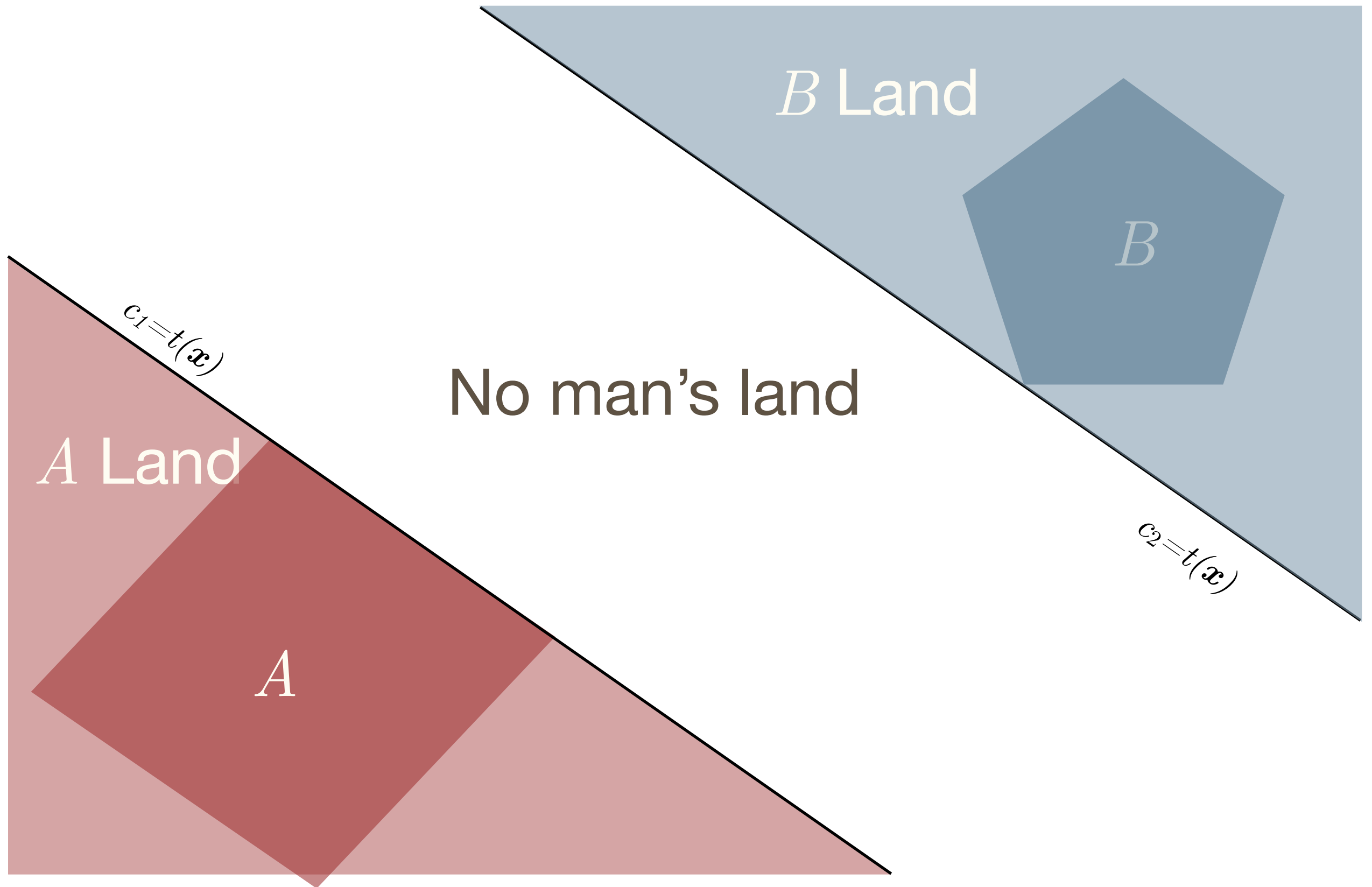
Interpolant Duality Visualized



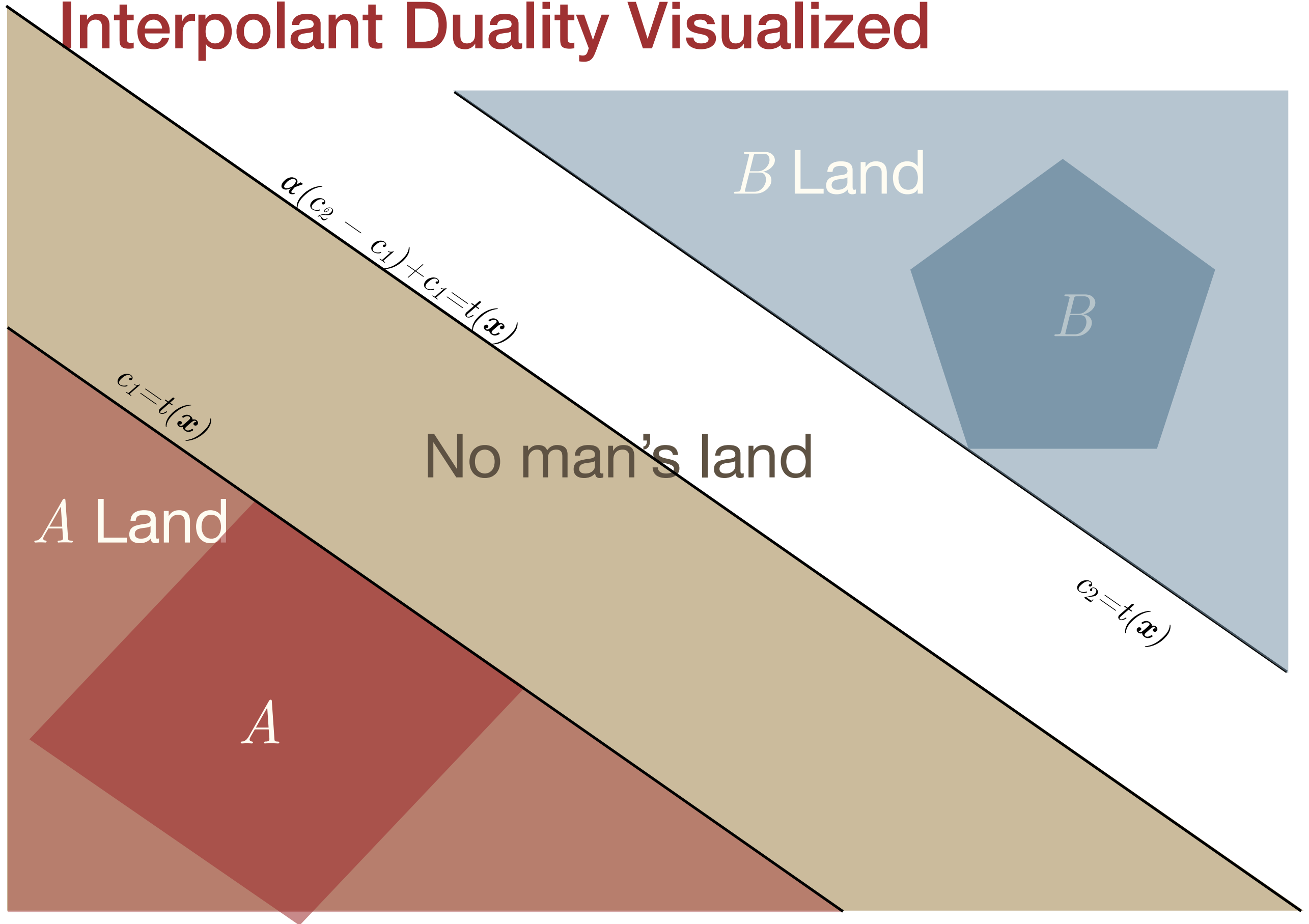
Interpolant Duality Visualized



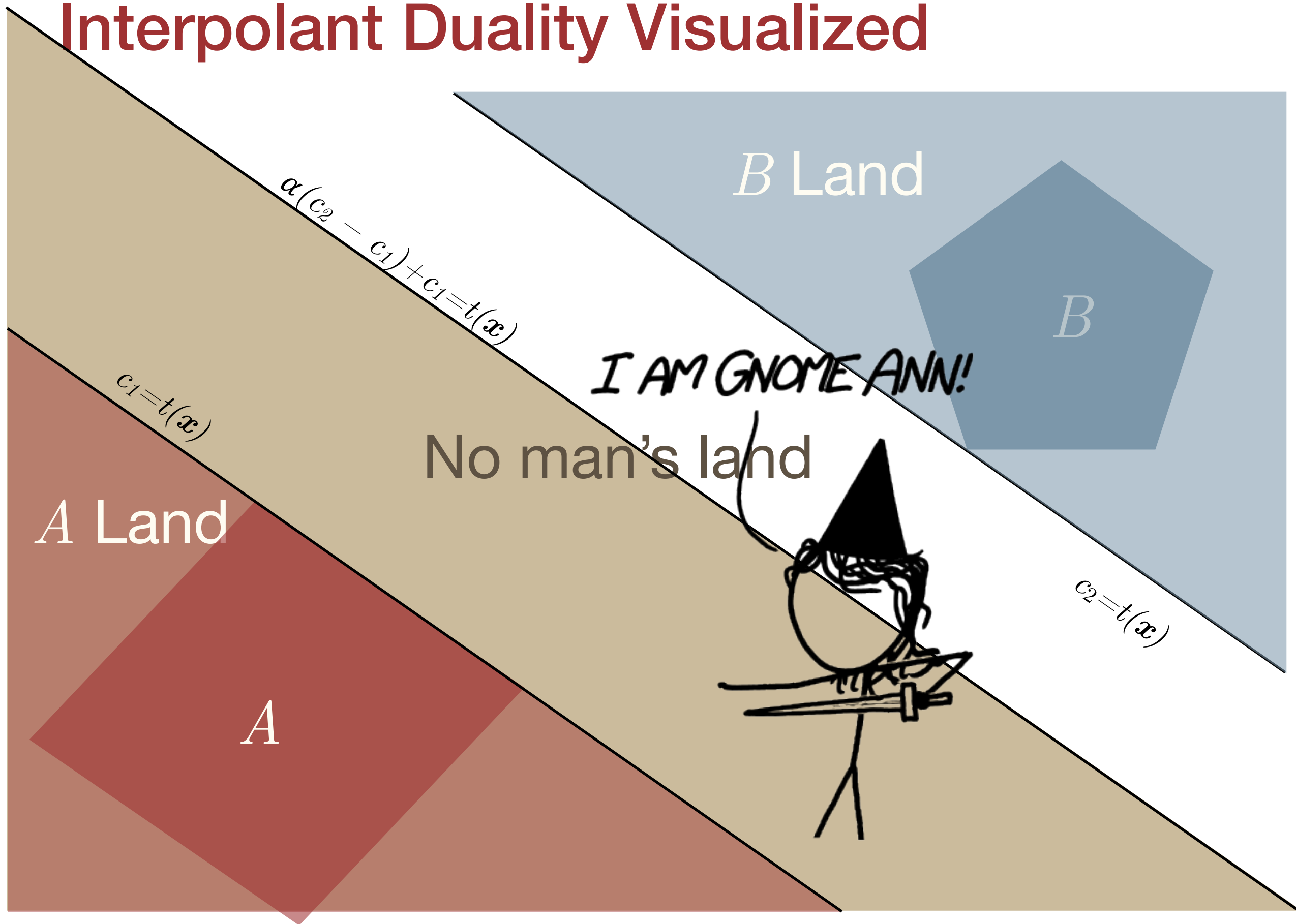
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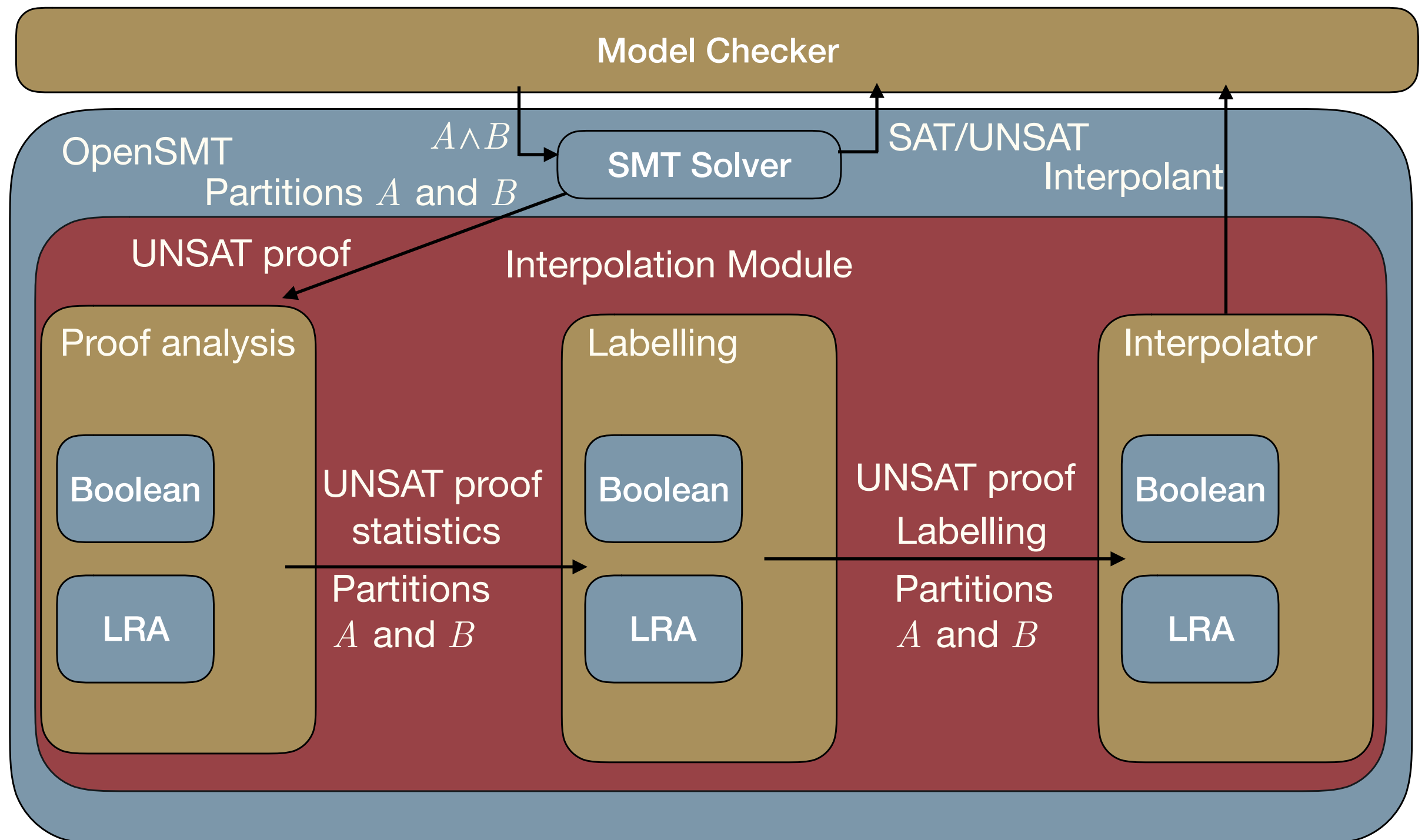


Interpolant Duality Visualized

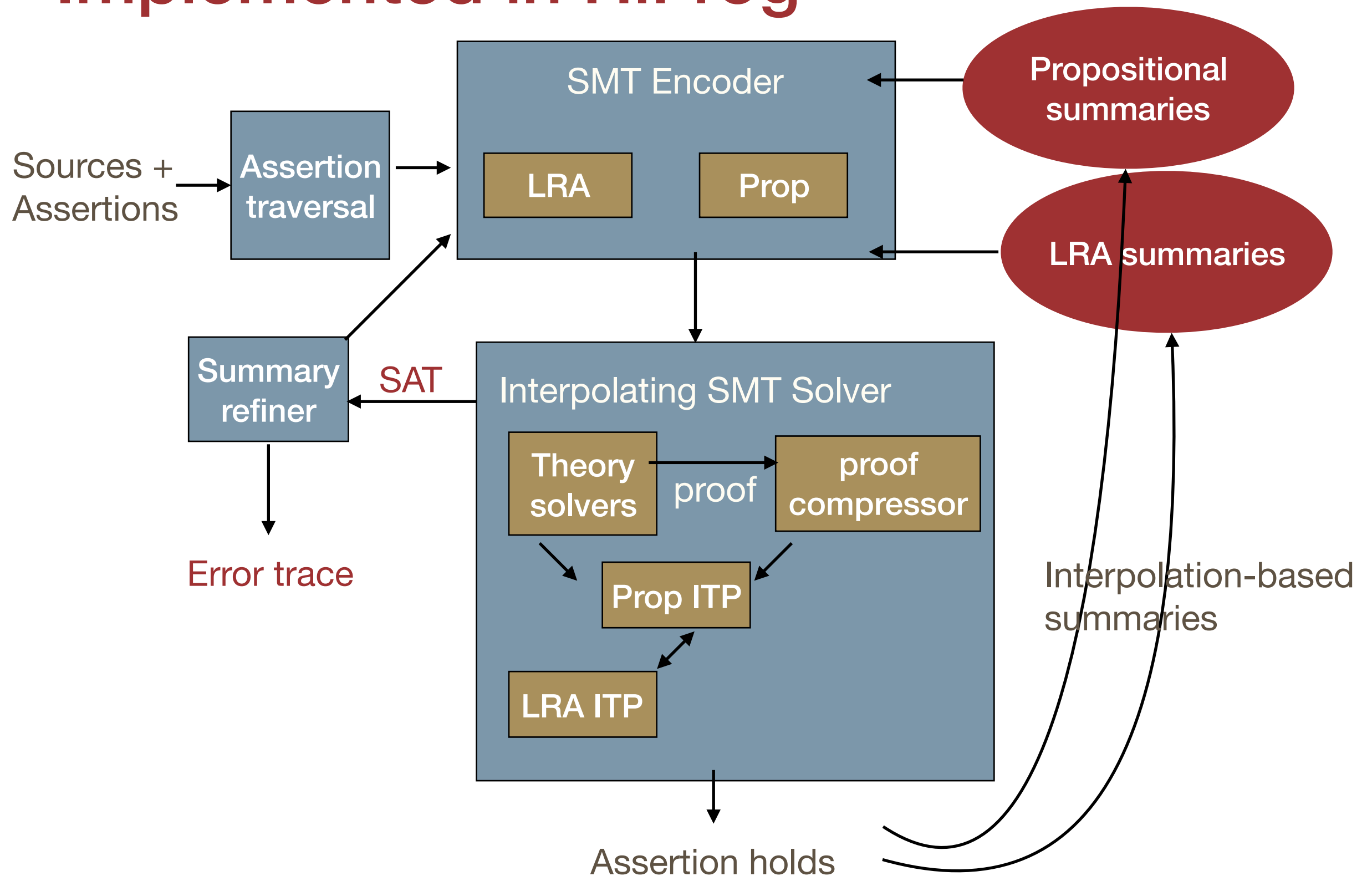


Experiments on SV-COMP and HiFrog

The Architecture Overview



Implemented in HiFrog



Results on SMT-LIB

Experiments with three LRA labelling functions:

Strong: the primal interpolant

Weak: the dual interpolant

$c = 0.5$: the interpolant between dual and primal

Experiments on HiFrog

ITP	floppy1	kbfiltr1	diskperf1	mem	disk	Σ
Strong	27100	5120	39900	25600	47600	145000
$c = 0.5$	25100	5120	39200	25100	41500	136000
Weak	24800	5380	39200	25600	64000	159000

Number of HiFrog refinements (fixed propositional ITP algorithm)

- The difference between minimum and maximum is $\sim 15\%$
- The $c = 0.5$ ITP provides the best results

Related Work

Nikolaj Bjørner, Arie Gurfinkel:

Property Directed Polyhedral Abstraction. VMCAI 2015

Pudlák:

Lower bounds for resolution and cutting plane proofs and monotone computations. Journal of Symbolic Logic 1997.

McMillan:

An Interpolating Theorem Prover. Theoretical Computer Science 2005.

D'Silva, Kroening, Purandare, and Weissenbacher:

Interpolant Strength. VMCAI 2010.

Albarghouthi, McMillan:

Beautiful interpolants. CAV 2013.

Dutertre, de Moura:

A fast linear-arithmetic solver for DPLL(T). Logical Methods in Computer Science 2012.

Alt, Hyvärinen, Asadi, and Sharygina:

Duality-Based Interpolation for Quantifier-Free Equalities and Uninterpreted Functions. FMCAD 2017.

Conclusions

LRA interpolation with controlled strength

Provides an infinite family of interpolants based on interpolation duality

Integrated into a model checker

Future work

Better heuristics for the labelling function

Apply to fix-point computations in other MC applications

Implementations available at

<http://verify.inf.usi.ch/hifrog>,

<http://verify.inf.usi.ch/opensmt>