

From Farkas' Lemma to Linear Programming: an Exercise in Diagrammatic Algebra

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Abstract

Farkas' lemma is a celebrated result on the solutions of systems of linear inequalities, which finds application pervasively in mathematics and computer science. In this work we show how to formulate and prove Farkas' lemma in diagrammatic polyhedral algebra, a sound and complete graphical calculus for polyhedra. Furthermore, we show how linear programs can be modeled within the calculus and how some famous duality results can be proved.

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1 Introduction

Farkas' lemma is a classical result on the solutions of systems of linear inequalities, which appears ubiquitously across various fields of Mathematics and Computer Science; more than a century after its introduction in [16, 17], it continues to receive attention and generate new lines of research [3, 10, 15, 22, 30, 25, 31, 4, 24, 28, 1]. Throughout the decades, different proofs have been given, and many variations have been proposed. The most established formulation asserts that, given an $m \times n$ matrix A , a vector $b \in \mathbb{R}^m$ and their transposes A^T and b^T , exactly one of the following two propositions is true.

- (a) $\exists x \in \mathbb{R}^n$ s.t. $x \geq 0$ and $Ax = b$ (b) $\exists y \in \mathbb{R}^m$ s.t. $A^T y \geq 0$ and $b^T y < 0$

Farkas' lemma finds application in a number of different scenarios, ranging from non-linear optimisation [28, 4] to the algebraic semantics of non-deterministic and probabilistic systems [23]. Most computer scientists first meet Farkas' lemma when studying duality theory in linear programming. A gentle introduction to this theory is provided by the *farmer problem*.

A farmer grows wheat and barley on a land of size l , with a provision f of fertilizer and p of pesticide. To grow one unit of wheat the farmer needs one unit of land, f_1 units of fertilizer and p_1 units of pesticide. Analogously, one unit of barley requires one unit of land, f_2 of fertilizer and p_2 of pesticide. The sell prices for wheat and barley are, respectively, s_1 and



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s_2 . By fixing x_1 to be the units of wheat and x_2 those of barley to be produced, the farmer should solve the following linear program to maximize the profit out of the production.

$$\max\{c \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mid x_1, x_2 \geq 0, A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \leq b\} \text{ where } c = (s_1 \ s_2), A = \begin{pmatrix} 1 & 1 \\ f_1 & f_2 \\ p_1 & p_2 \end{pmatrix}, b = \begin{pmatrix} l \\ f \\ p \end{pmatrix}$$

Now assume that a planning board needs to establish prices for land, fertilizer and pesticide. The board's job is to minimize the cost of production while assuring some profit to the farmer. To do so, it is sufficient to solve the following program where A , b and c are as above.

$$\min\{b^T \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \mid y_1, y_2, y_3 \geq 0, A^T \begin{pmatrix} y_1 \\ y_2 \\ y_3 \end{pmatrix} \leq c^T\}$$

The problem of the farmer and the one of the board are a typical example of a pair of dual problems. A result in duality theory (which makes the relevance of Farkas' lemma apparent) is that, if a problem has unbounded solution, then its dual has no solution. Most importantly, when a problem and its dual have finite solutions, then these solutions coincide. In the example above, the minimum cost of the production and the maximum profit of the farmer should then be equal.

In this paper we revisit Farkas' lemma and duality results in linear programming through the lens of *string diagrams*.

String diagrams are a graphical syntax for representing arrows of symmetric monoidal categories [33]. In recent years, increasingly they have been adopted as a formal language to study component-based systems across different fields of science [12, 2, 18, 19, 21, 29, 32] using the compositional methods that are typical of programming language semantics. One striking property of this approach is that, even though string diagrams have an appealing graphical representation, they are completely formal syntactic objects. Furthermore, they may receive semantics interpretation in some mathematical domain (such as functions, relations, matrices, subspaces, etc.) and many results have been provided on how equational theories of string diagrams are able to *axiomatise* semantic equality over these domains, see e.g. [6, 8, 36, 37, 2, 7]. Such a *complete* equational theory yields a powerful pictorial calculus to reason algebraically about system behaviour, for instance in concurrency [6, 11], control [9, 2] and quantum theory [13].

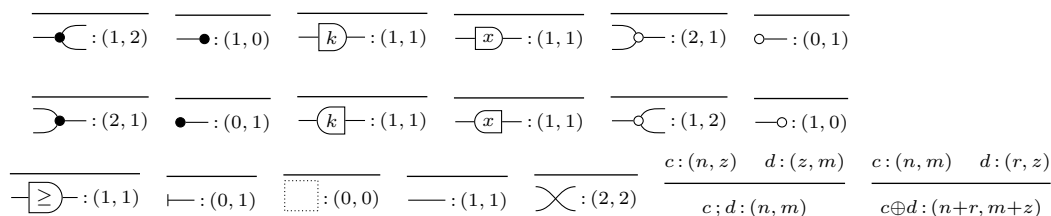
The core of the calculus that we exploit in this paper is the theory of Interacting Hopf Algebras [36, 8, 2], originally introduced to reason about the behaviour of signal flow graphs [34]. Such theory has been extended first in [7] to study non-passive electrical network and concurrent connectors [11], and then in [5], for studying continuous Petri nets [14]. The latter extension, called *diagrammatic polyhedral algebra*, provides a sound and complete calculus which is able to express exactly *polyhedra*. We claim this is the proper string diagrammatic setting to express Farkas' lemma and duality in linear programming.

In diagrammatic polyhedral algebra, recalled in Section 2, different entities of traditional algebra, like vectors, matrices and subsets $C \subseteq \mathbb{R}^n$ are all regarded as relations amongst vectors spaces. Starting from few primitive relations (depicted as wires and gates of circuits), one can syntactically construct all polyhedra by means of relational composition and cartesian product (graphically rendered as horizontal and vertical juxtaposition). It is exactly this linguistic aspect the main novelty of our proof of Farkas' lemma: statements about existence of solutions, like (a) and (b) above, translate into equations amongst terms of the string diagrammatic syntax; proofs are symbolic manipulation of diagrams, whose soundness is

guaranteed by the axiomatisation. Moreover, compositionality allows to break complex notions into simple inductive definitions on the sets of primitive relations. For instance, the *polar operator* which is given inductively in Section 3, captures the notions of polar and dual cone that are defined in the traditional language by mean of universal quantifications.

In the context of diagrammatic polyhedral algebra, the proof of Farkas' lemma becomes straightforward: using a basic observation, named the *lemma of alternatives* in Section 4, the proof—in Section 5—reduces to compute the polar operator over a certain string diagram.

The final part of our work (Section 6) is dedicated to duality in linear programming. Interestingly, diagrammatic polyhedral algebra allows to prove various duality theorems in a rather different way than those found in traditional textbooks (see e.g. [35]). In the classical approach, one first needs to massage the dual problems to bring them into an appropriate shape, and then prove, in sequence, a weak and a strong duality theorems. Our proof method instead is based on a general principle (Theorem 23) that, independently from the shape of the problem at hand, allows to prove all the results at once. Curiously, our proof does not rely on Farkas' lemma: rather both the duality theorems and Farkas' lemma stem from general results encoded in the axiomatisation of diagrammatic polyhedral algebra.



■ **Figure 1** Sort inference rules.

2 Diagrammatic Polyhedral Algebra

This section presents a calculus of string diagrams for reasoning about polyhedra, which we will later use to prove Farkas' Lemma and the duality theorems for linear programming. The calculus was first introduced in [5], to which we refer for a more detailed exposition.

We fix an *ordered field* k , i.e. a field equipped with a total order \geq such that for all $i, j, k \in k$: (a) if $i \geq j$, then $i + k \geq j + k$; (b) if $i \geq 0$ and $j \geq 0$, then $i \cdot j \geq 0$. The syntax of the calculus is given by the following context free grammar, where k ranges over k .

$$c ::= \text{---}\bullet \mid \text{---}\bullet\text{---} \mid \text{---}\boxed{k}\text{---} \mid \text{---}\text{---} \mid \text{---}\circ \mid \quad (1)$$

$$\bullet\text{---} \mid \text{---}\bullet\text{---} \mid \text{---}\boxed{k}\text{---} \mid \text{---}\text{---} \mid \text{---}\circ \mid \quad (2)$$

$$\text{---}\boxed{\geq}\text{---} \mid \quad (3)$$

$$\text{---} \mid \quad (4)$$

$$\text{---}\square\text{---} \mid \text{---} \mid \text{---}\times\text{---} \mid c;c \mid c \oplus c \quad (5)$$

We shall consider only terms that are *sortable*, i.e. that one may associate with a pair (n, m) of natural numbers $n, m \in \mathbb{N}$ using the rules in Figure 1.

The above syntax specification purposefully uses a graphical rendering of the components. As customary for string diagrams, we will render composition via $;$ and \oplus graphically by

■ The symmetries $\sigma_{n,m} : n + m \rightarrow m + n$ are the relations

$$\left\{ \left(\begin{pmatrix} u \\ v \end{pmatrix}, \begin{pmatrix} v \\ u \end{pmatrix} \right) \mid u \in k^n, v \in k^m \right\}$$

We can now formally define the semantic interpretation as a *prop morphism* (an identity-on-objects symmetric monoidal functor) $\llbracket \cdot \rrbracket : \text{PDiag} \rightarrow \text{Rel}_k$. For the generators in (1), $\llbracket \cdot \rrbracket$ is

$$\begin{aligned} \llbracket \text{---} \circlearrowleft \rrbracket &= \left\{ \left(x, \begin{pmatrix} x \\ x \end{pmatrix} \right) \mid x \in k \right\} & \llbracket \text{---} \circlearrowright \rrbracket &= \left\{ \left(\begin{pmatrix} x \\ y \end{pmatrix}, x + y \right) \mid x, y \in k \right\} \\ \llbracket \text{---} \bullet \rrbracket &= \{ (x, \bullet) \mid x \in k \} & \llbracket \text{---} \circ \rrbracket &= \{ (\bullet, 0) \} & \llbracket \text{---} \overline{k} \rrbracket &= \{ (x, k \cdot x) \mid x \in k \} \end{aligned} \quad (6)$$

and, symmetrically, for the generators in (2). For instance, $\llbracket \text{---} \overline{k} \rrbracket = \{ (k \cdot x, x) \mid x \in k \}$. For the generators in (3) and (4), the semantics is defined, respectively, as $\llbracket \text{---} \geq \rrbracket = \{ (x, y) \mid x, y \in k, x \geq y \}$ and $\llbracket \text{---} \rrbracket = \{ (\bullet, 1) \}$. The semantics of the identities, symmetries and compositions – in (5) – is given by the *functoriality* of $\llbracket \cdot \rrbracket$, e.g., $\llbracket c; d \rrbracket = \llbracket c \rrbracket ; \llbracket d \rrbracket$ and $\llbracket \text{---} \rrbracket = \llbracket id_0 \rrbracket = \{ (\bullet, \bullet) \}$. Above we used \bullet for the unique element of the vector space k^0 .

► **Example 3.** Two string diagrams will play a special role in our exposition: $\bullet \circlearrowleft$ and $\circlearrowright \bullet$. By definition of $\llbracket \cdot \rrbracket$, note that their semantics forces the two ports on the right (resp. left) to carry the same value, thus acting as a left (right) feedback.

$$\llbracket \bullet \circlearrowleft \rrbracket = \left\{ \left(\bullet, \begin{pmatrix} x \\ x \end{pmatrix} \right) \mid x \in k \right\} \quad \llbracket \circlearrowright \bullet \rrbracket = \left\{ \left(\begin{pmatrix} x \\ x \end{pmatrix}, \bullet \right) \mid x \in k \right\}$$

We can use these feedback diagrams to arbitrarily move wires from left to right. For instance

$$\text{---} \leq \text{---} := \text{---} \geq \text{---} \quad \text{---} \dashv := \text{---} \bullet$$

As expected, $\llbracket \text{---} \leq \rrbracket = \{ (y, x) \mid x, y \in k, x \geq y \}$ and $\llbracket \text{---} \dashv \rrbracket = \{ (1, \bullet) \}$.

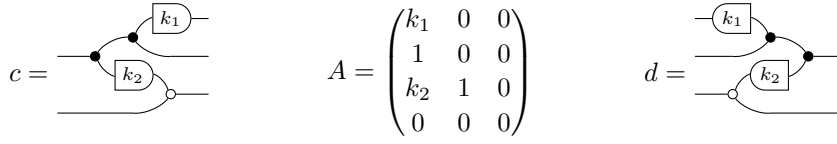
In [5] it is shown that diagrams of PDiag can express, amongst all the relations $R \subseteq k^n \times k^m$, exactly all those that are *polyhedra*, cf. Example 7 below. Moreover, it is worth recalling that fragments of PDiag also characterise well-known classes of relational objects, as indicated in the table below (see [36, 5] for an overview of these results).

prop	syntax	semantics
$\text{MDiag}^{\rightarrow}$	(1), (5)	matrices
$\text{MDiag}^{\leftarrow}$	(2), (5)	reversed matrices
LDiag	(1), (2), (5)	linear relations (sub-spaces)
PCDiag	(1), (2), (3), (5)	polyhedral cones
PDiag	(1), (2), (3), (4), (5)	polyhedra

(7)

For instance, the arrows of PDiag, which are only built from the components in (1) and (5), form a sub-prop of PDiag, denoted by $\text{MDiag}^{\rightarrow}$, and characterise k -matrices – in terms of the semantics functor $\llbracket \cdot \rrbracket : \text{PDiag} \rightarrow \text{Rel}_k$, they denote precisely the relations of the form $\{ (x, Ax) \mid x \in k^p \}$ for some matrix A . Similarly $\text{MDiag}^{\leftarrow}$, LDiag and PCDiag are the sub-props of PDiag of arrows built from the generators specified in (7). Hereafter we illustrate some examples of these fragments, and the corresponding semantic characterisation.

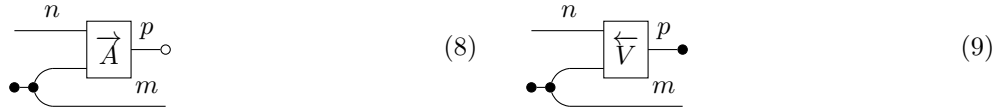
► **Example 4** ((Reversed) Matrices). As mentioned, diagrams $c: n \rightarrow m$ in $\text{MDiag}^{\rightarrow}$ denote precisely the $m \times n$ matrices (see [36] for all details). Consider for instance, the diagram $c: 3 \rightarrow 4$ and its representation as a 4×3 matrix. Note that $A_{ij} = k$ whenever k is the scalar encountered on the path from the i th port to the j th port. If there is no path, then $A_{ij} = 0$. It is easy to check that $\llbracket c \rrbracket = \{(x, y) \in \mathbb{k}^3 \times \mathbb{k}^4 \mid y = Ax\}$.



Dually, diagrams in $\text{MDiag}^{\leftarrow}$ are “reversed” matrices: inputs on the right and outputs on the left. For instance $d: 4 \rightarrow 3$ again encodes A , but its semantics is $\llbracket d \rrbracket = \{(y, x) \in \mathbb{k}^4 \times \mathbb{k}^3 \mid y = Ax\}$.

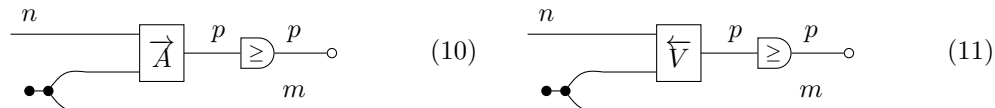
Hereafter we will use \overrightarrow{A}^m and \overleftarrow{A}^n for some diagrams of $\text{MDiag}^{\rightarrow}$ and, respectively $\text{MDiag}^{\leftarrow}$, corresponding to some $m \times n$ matrix A . For matrices of type $m \times 1$ and $1 \times n$ we will use lower case letters, usually b and c respectively. It is worth remarking that while $m \times 1$ matrices and vectors in \mathbb{k}^m have the same representation in the traditional notation, in PDiag , they are presented as \overrightarrow{b}^m and \overleftarrow{b}^m . Indeed, the semantics of the former is $\{(k, bk) \in \mathbb{k}^1 \times \mathbb{k}^m \mid k \in \mathbb{k}\}$, while the semantics of the latter is $\{(\bullet, b) \in \mathbb{k}^0 \times \mathbb{k}^m\}$.

► **Example 5** (Linear Relations). Consider the following diagrams in LDiag .



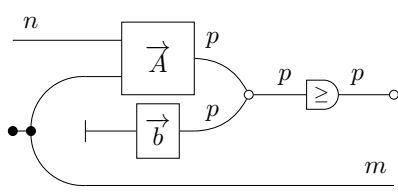
It is easy to check that the semantics of (8) is the set $\{(x, y) \in \mathbb{k}^n \times \mathbb{k}^m \mid A \begin{pmatrix} x \\ y \end{pmatrix} = 0\}$, that is the set of solutions of some system of linear equations. Such system has p rows in $n + m$ variables: n variables stand on the left and m variables on the right. This means that $\llbracket (8) \rrbracket$ is a sub-vector space of $\mathbb{k}^n \times \mathbb{k}^m$, namely a *linear relation*. The semantics of (9) is $\{(x, y) \in \mathbb{k}^n \times \mathbb{k}^m \mid \exists z \in \mathbb{k}^p \text{ s.t. } \begin{pmatrix} x \\ y \end{pmatrix} = Vz\}$, that is the linear hull of the set of column vectors of the matrix V , or in other words the subspace generated by V . Recall that any subspace can be represented both in the form of a system of linear equations and in the form of a set of generating vectors. Indeed, diagrams (8) and (9) represents two normal forms for the diagrams in LDiag .

► **Example 6** (Polyhedral cones). Consider the following diagrams in PCDiag

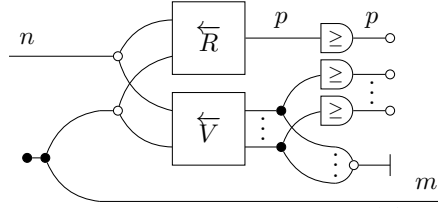


with semantics $\{(x, y) \in \mathbb{k}^n \times \mathbb{k}^m \mid A \begin{pmatrix} x \\ y \end{pmatrix} \geq 0\}$ and $\{(x, y) \in \mathbb{k}^n \times \mathbb{k}^m \mid \exists z \in \mathbb{k}^p \text{ s.t. } \begin{pmatrix} x \\ y \end{pmatrix} = Vz, z \geq 0\}$, respectively. The semantics of (10) is thus the set of solutions of a systems of linear *inequalities*, namely a *polyhedral cone*, while the semantics of (11) is the conic hull of V (seen as a set of column vectors). Similarly to Example 5, diagrams in (10) and (11) can be regarded as two normal forms for diagrams in PCDiag .

► **Example 7** (Polyhedra). Consider the following diagrams in PDiag.



(12)



(13)

It is easy to check that the semantics of (12) is the relation $\{(x, y) \in \mathbb{k}^n \times \mathbb{k}^m \mid A \begin{pmatrix} x \\ y \end{pmatrix} + b \geq 0\}$ and thus the representation of a *polyhedron* as the set of solutions of a system of affine inequalities. The semantics of (13) is the relation $\{(x, y) \in \mathbb{k}^n \times \mathbb{k}^m \mid \exists z \in \mathbb{k}^p, w \in \mathbb{k}^o \text{ s.t. } z \geq 0, w \geq 0, \sum w_i = 1, Rz + Vw = \begin{pmatrix} x \\ y \end{pmatrix}\}$ and thus a *vertex* representation of a polyhedron. In other words, $\llbracket (13) \rrbracket$ is Minkowsky sum of the conic hull of $R, \{\begin{pmatrix} x \\ y \end{pmatrix} \mid \exists z \in \mathbb{k}^p, \text{ s.t. } z \geq 0, Rz = \begin{pmatrix} x \\ y \end{pmatrix}\}$, and of the *convex hull* of $V, \{\begin{pmatrix} x \\ y \end{pmatrix} \mid \exists w \in \mathbb{k}^o \text{ s.t. } w \geq 0, \sum w_i = 1, Vw = \begin{pmatrix} x \\ y \end{pmatrix}\}$.

The functor $\llbracket \cdot \rrbracket : \text{PDiag} \rightarrow \text{Rel}_{\mathbb{k}}$ is not *faithful*: two different string diagrams may denote the same relation. However, PDiag can be equipped with a *sound and complete axiomatisation*, meaning an equational theory making two diagrams c and d equal precisely when $\llbracket c \rrbracket = \llbracket d \rrbracket$. Such axiomatisation, called *Polyhedral Algebra* (\mathbb{PA}) is illustrated in Figure 2, where we write $l = r$ for the two inequalities $l \sqsubseteq r$ and $r \sqsubseteq l$. In order to state the completeness theorem, we define $\sqsubseteq^{\mathbb{PA}}$ as the smallest precongruence containing all the pairs (c, d) such that $c \sqsubseteq d$ appears in the Figure 2. In other words, $\sqsubseteq^{\mathbb{PA}}$ is the smallest relation containing \sqsubseteq which is closed by reflexivity, transitivity, composition ; and monoidal product \otimes . Finally, we write $c \stackrel{\mathbb{PA}}{=} d$ iff $c \sqsubseteq^{\mathbb{PA}} d$ and $d \sqsubseteq^{\mathbb{PA}} c$.

► **Theorem 8** (From [5]). *For all diagrams c, d in PDiag, $\llbracket c \rrbracket \subseteq \llbracket d \rrbracket$ if and only if $c \stackrel{\mathbb{PA}}{=} d$.*

Here are some interesting consequences of the theory \mathbb{PA} , where we use \blacksquare for $\boxed{-1}$.

$$\circ \text{---} \boxed{\geq} \text{---} \circ \stackrel{\mathbb{PA}}{=} \circ \text{---} \circ$$

(14)

$$\blacksquare \text{---} \circ \text{---} \circ \stackrel{\mathbb{PA}}{=} \circ \text{---} \blacksquare$$

(15)

$$\blacksquare \text{---} \boxed{A} \text{---} \blacksquare \stackrel{\mathbb{PA}}{=} \blacksquare \text{---} \boxed{A} \text{---} \blacksquare \text{ for any } A \text{ in LDiag}$$

(16)

$$\blacksquare \text{---} \blacksquare \stackrel{\mathbb{PA}}{=} \text{---}$$

(17)

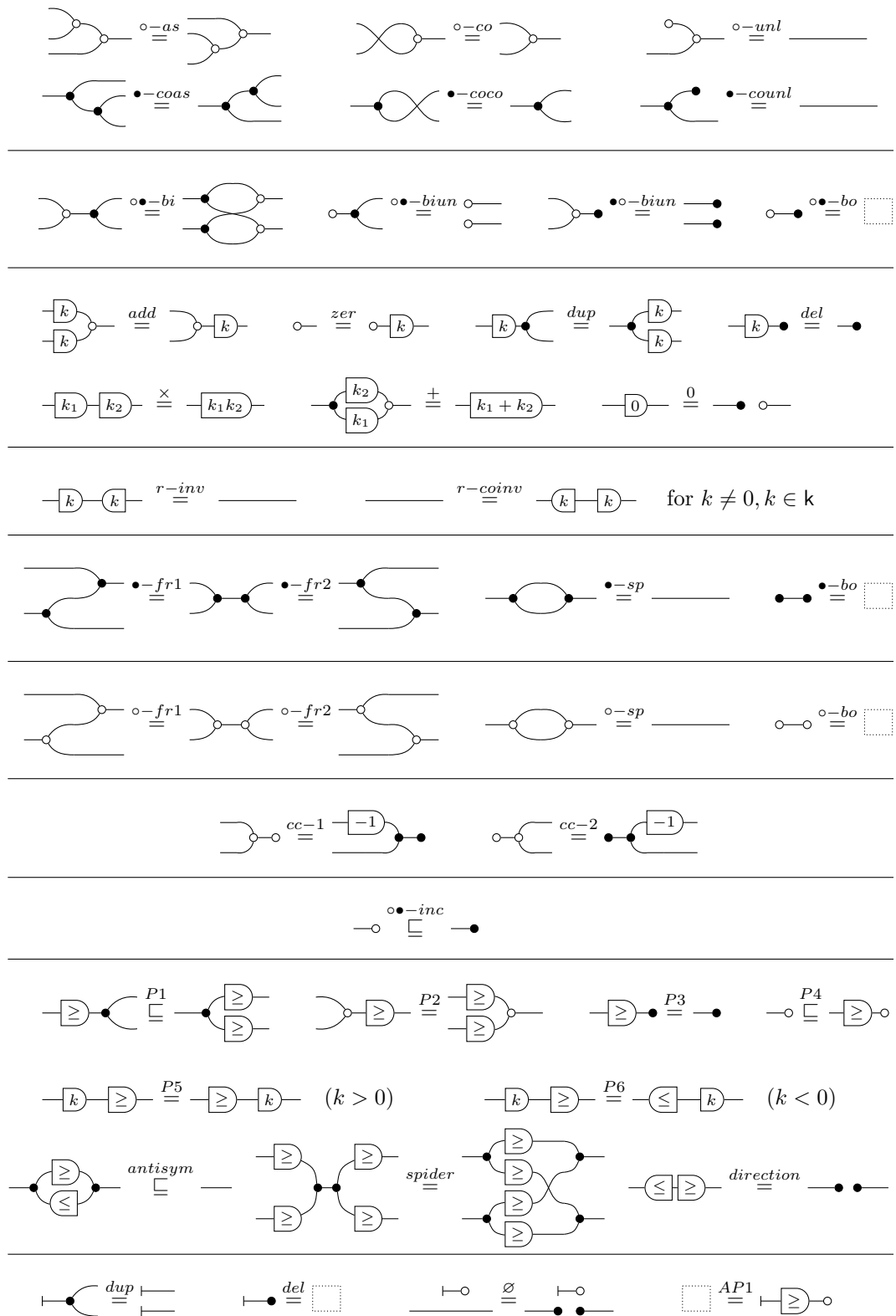
Theorem 8 implies that equivalences like (14), (15) and (17) may be also proved by purely graphical means, using derivations involving the axioms of \mathbb{PA} , without resorting to the semantic interpretation $\llbracket \cdot \rrbracket$. The proofs of more sophisticated statements, as (16), involve axioms in combination with other proof techniques, e.g., induction.

The following is an example of derivation proving (14).

$$\circ \text{---} \circ \stackrel{P4}{\sqsubseteq} \circ \text{---} \boxed{\geq} \text{---} \circ \quad \text{and} \quad \circ \text{---} \circ \stackrel{antisym}{\sqsupseteq} \circ \text{---} \begin{matrix} \boxed{\geq} \\ \boxed{\leq} \end{matrix} \text{---} \circ \stackrel{dup}{=} \circ \text{---} \boxed{\geq} \text{---} \circ \stackrel{AP1}{=} \circ \text{---} \boxed{\geq} \text{---} \circ \quad (18)$$

Note that in (18) we used a version of axioms $P4$, dup and $AP1$ where diagrams are “rotated over the y axis”. We formalise such a notion, in a way that justifies this use.

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■ Figure 2 Axioms of $\mathbb{P}A_k$.

► **Definition 9.** The prop morphism $\cdot^{op}: \text{PDiag}^{op} \rightarrow \text{PDiag}$ is inductively defined as:

$$\begin{array}{ccccccc}
 \begin{array}{c} \bullet \\ \circlearrowleft \end{array}^{op} = \begin{array}{c} \bullet \\ \circlearrowright \end{array} & \begin{array}{c} \bullet \\ \bullet \end{array}^{op} = \begin{array}{c} \bullet \\ \bullet \end{array} & \begin{array}{c} \boxed{k} \\ \text{---} \end{array}^{op} = \begin{array}{c} \boxed{k} \\ \text{---} \end{array} & \begin{array}{c} \circlearrowleft \\ \circlearrowright \end{array}^{op} = \begin{array}{c} \circlearrowright \\ \circlearrowleft \end{array} & \begin{array}{c} \circ \\ \text{---} \end{array}^{op} = \begin{array}{c} \circ \\ \text{---} \end{array} \\
 \begin{array}{c} \circlearrowright \\ \bullet \end{array}^{op} = \begin{array}{c} \circlearrowleft \\ \bullet \end{array} & \begin{array}{c} \bullet \\ \bullet \end{array}^{op} = \begin{array}{c} \bullet \\ \bullet \end{array} & \begin{array}{c} \boxed{k} \\ \text{---} \end{array}^{op} = \begin{array}{c} \boxed{k} \\ \text{---} \end{array} & \begin{array}{c} \circlearrowright \\ \circlearrowleft \end{array}^{op} = \begin{array}{c} \circlearrowleft \\ \circlearrowright \end{array} & \begin{array}{c} \circ \\ \text{---} \end{array}^{op} = \begin{array}{c} \circ \\ \text{---} \end{array} \\
 \begin{array}{c} \boxed{\geq} \\ \text{---} \end{array}^{op} = \begin{array}{c} \boxed{\leq} \\ \text{---} \end{array} & \text{---}^{op} = \text{---} & \begin{array}{c} \square \\ \text{---} \end{array}^{op} = \begin{array}{c} \square \\ \text{---} \end{array} & \text{---}^{op} = \text{---} & \begin{array}{c} \times \\ \text{---} \end{array}^{op} = \begin{array}{c} \times \\ \text{---} \end{array} \\
 (c; d)^{op} = d^{op}; c^{op} & & (c \oplus d)^{op} = c^{op} \oplus d^{op} & &
 \end{array}$$

Observe that \cdot^{op} is contravariant: it maps a diagram $c: n \rightarrow m$ into $c^{op}: m \rightarrow n$ which is graphically rendered as the *mirror image* of c : for instance, referring to Example 4, $c^{op} = d$. By exploiting the inductive definition, one can prove that the following hold.

$$\left(\begin{array}{c} n \\ \boxed{c} \\ m \end{array} \right)^{op} \stackrel{\mathbb{P}\mathbb{A}}{=} \begin{array}{c} n \\ \bullet \\ \bullet \\ \boxed{c} \\ \bullet \\ m \end{array} \quad (19) \quad \left(\begin{array}{c} n \\ \boxed{A} \\ m \end{array} \right)^{op} \stackrel{\mathbb{P}\mathbb{A}}{=} m \boxed{A} n \quad (20)$$

$$\text{if } c \stackrel{\mathbb{P}\mathbb{A}}{\sqsubseteq} d \text{ then } c^{op} \stackrel{\mathbb{P}\mathbb{A}}{\sqsubseteq} d^{op} \quad (21) \quad (c^{op})^{op} \stackrel{\mathbb{P}\mathbb{A}}{=} c \quad (22)$$

Equation (19) states that $\llbracket c^{op} \rrbracket$ is exactly the *opposite relation* of $\llbracket c \rrbracket$, i.e., $\llbracket c^{op} \rrbracket = \{(y, x) \in k^m \times k^n \mid (x, y) \in \llbracket c \rrbracket\}$. In particular, by (20), any diagram in $\text{MDiag}^{\rightarrow}$ representing a matrix A is mapped into a diagram in $\text{MDiag}^{\leftarrow}$ representing the same matrix (see Example 4). Thanks to (21) and (22), one has that $c \stackrel{\mathbb{P}\mathbb{A}}{\sqsubseteq} d$ iff $c^{op} \stackrel{\mathbb{P}\mathbb{A}}{\sqsubseteq} d^{op}$. Therefore, each of the axioms in Figure 2 and each of the laws that we prove in this text can be read both as $c \stackrel{\mathbb{P}\mathbb{A}}{\sqsubseteq} d$ and as $c^{op} \stackrel{\mathbb{P}\mathbb{A}}{\sqsubseteq} d^{op}$. For instance, by (15) we also know that $\begin{array}{c} \bullet \\ \circlearrowleft \end{array} \stackrel{\mathbb{P}\mathbb{A}}{=} \begin{array}{c} \bullet \\ \bullet \end{array}$. Like in (18), in our derivations we will always use this property implicitly.

3 The polar operator

When reasoning about cones $C \subseteq k^n$ in convex algebra, an important role is played by the notions of *polar* and *dual* cone:

$$\text{polar}(C) = \{b \in k^n \mid \forall x \in C, b^T x \leq 0\} \quad \text{dual}(C) = \{b \in k^n \mid \forall x \in C, b^T x \geq 0\}$$

As these concepts will also be relevant to our developments, we now study how they are expressible in PCDiag. The fundamental ingredient is the *polar operator* from [5]:

► **Definition 10.** The prop morphism $\cdot^\circ: \text{PCDiag} \rightarrow \text{PCDiag}$ is inductively defined as:

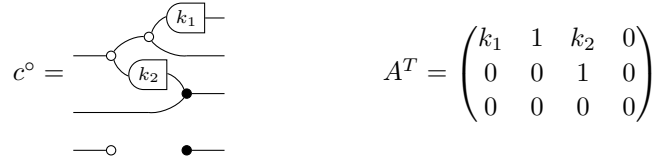
$$\begin{array}{ccccccc}
 \begin{array}{c} \bullet \\ \circlearrowleft \end{array}^\circ = \begin{array}{c} \bullet \\ \circlearrowright \end{array} & \begin{array}{c} \bullet \\ \bullet \end{array}^\circ = \begin{array}{c} \bullet \\ \bullet \end{array} & \begin{array}{c} \boxed{k} \\ \text{---} \end{array}^\circ = \begin{array}{c} \boxed{k} \\ \text{---} \end{array} & \begin{array}{c} \circlearrowleft \\ \circlearrowright \end{array}^\circ = \begin{array}{c} \circlearrowright \\ \circlearrowleft \end{array} & \begin{array}{c} \circ \\ \text{---} \end{array}^\circ = \begin{array}{c} \bullet \\ \text{---} \end{array} \\
 \begin{array}{c} \circlearrowright \\ \bullet \end{array}^\circ = \begin{array}{c} \circlearrowleft \\ \bullet \end{array} & \begin{array}{c} \bullet \\ \bullet \end{array}^\circ = \begin{array}{c} \bullet \\ \bullet \end{array} & \begin{array}{c} \boxed{k} \\ \text{---} \end{array}^\circ = \begin{array}{c} \boxed{k} \\ \text{---} \end{array} & \begin{array}{c} \circlearrowright \\ \circlearrowleft \end{array}^\circ = \begin{array}{c} \circlearrowleft \\ \circlearrowright \end{array} & \begin{array}{c} \circ \\ \text{---} \end{array}^\circ = \begin{array}{c} \bullet \\ \text{---} \end{array} \\
 \begin{array}{c} \square \\ \text{---} \end{array}^\circ = \begin{array}{c} \square \\ \text{---} \end{array} & \text{---}^\circ = \text{---} & (c \oplus d)^\circ = c^\circ \oplus d^\circ & (c; d)^\circ = c^\circ; d^\circ & \begin{array}{c} \times \\ \text{---} \end{array}^\circ = \begin{array}{c} \times \\ \text{---} \end{array} \\
 \begin{array}{c} \boxed{\geq} \\ \text{---} \end{array}^\circ = \begin{array}{c} \bullet \\ \bullet \\ \boxed{\geq} \\ \bullet \end{array} & & & &
 \end{array}$$

The polar operator subsumes both the concept of dual and polar cone. This can be made precise via the following proposition, whose proof we defer to the end of the next section.

► **Proposition 11.** Let $C \subseteq k^n$ be a polyhedral cone. Let $c: 0 \rightarrow n$ and $d: n \rightarrow 0$ be such that $\llbracket c \rrbracket = \{(\bullet, x) \mid x \in C\}$ and $\llbracket d \rrbracket = \{(x, \bullet) \mid x \in C\}$. Then $\llbracket c^\circ \rrbracket = \{(\bullet, b) \mid b \in \text{polar}(C)\}$ and $\llbracket d^\circ \rrbracket = \{(b, \bullet) \mid b \in \text{dual}(C)\}$.

Note that Proposition 11 uses two representations of the cone C as a string diagram, one of type $0 \rightarrow n$ and the other of type $n \rightarrow 0$. Depending on which one we pick, one obtains the polar or the dual of C . Another interesting departure from the traditional approaches is that the polar/dual cone is now specified *inductively* on the structure of the string diagram, following Definition 10. We now provide some properties and examples of the polar operator. First we observe how it behaves on string diagrams representing matrices.

► **Example 12.** Consider the matrix A in Example 4 and its encoding as the diagram $c: 3 \rightarrow 4$ in $\text{MDiag}^{\rightarrow}$. Applying the polar operator on c yields the following diagram in $\text{MDiag}^{\leftarrow}$



representing the transpose A^T of the matrix A . Indeed, $\llbracket c^\circ \rrbracket = \{(x, y) \in \mathbb{k}^3 \times \mathbb{k}^4 \mid A^T y = x\}$. This is an instance of a more general phenomenon: when applied to matrices (i.e. string diagrams of $\text{MDiag}^{\rightarrow}$), the polar operator yields their transpose matrix, represented by a string diagram in $\text{MDiag}^{\leftarrow}$ (and thus to be read “right-to-left”).

► **Lemma 13** (From [36]). For all $\boxed{\overrightarrow{A}}: n \rightarrow m$ in $\text{MDiag}^{\rightarrow}$, the following holds

$$\boxed{\overrightarrow{A}}^\circ \stackrel{\text{PA}}{=} \boxed{\overleftarrow{A^T}}.$$

► **Proposition 14** (From [5]). For all diagrams $c, d: n \rightarrow m$ in PCDiag , the following hold

1. if $c \stackrel{\text{PA}}{\sqsubseteq} d$ then $(d)^\circ \stackrel{\text{PA}}{\sqsubseteq} (c)^\circ$;
2. $(c)^\circ \stackrel{\text{PA}}{=} c$.

The first item of the above proposition informs us that if $c \stackrel{\text{PA}}{=} d$ then one can safely conclude that $c^\circ \stackrel{\text{PA}}{=} d^\circ$. Viceversa, if $c^\circ \stackrel{\text{PA}}{=} d^\circ$, by the second item, $c \stackrel{\text{PA}}{=} d$. The next lemma illustrates the interaction of the polar operator with \cdot^{op} (see Definition 9).

► **Lemma 15.** For all $c: n \rightarrow m$ in PCDiag , the following holds $(c^{op})^\circ \stackrel{\text{PA}}{=} m \blacksquare^m; (c^\circ)^{op}; n \blacksquare^n$.

Proof.

$$(c^{op})^\circ \stackrel{(19)}{=} \left(\text{diagram of } c \text{ with inputs and outputs swapped} \right)^\circ \stackrel{(15)}{=} \text{diagram of } c^\circ \text{ with inputs and outputs swapped} \stackrel{(19)}{=} m \blacksquare^m; (c^\circ)^{op}; n \blacksquare^n \quad \blacktriangleleft$$

► **Example 16.** The diagrams $\circ \boxed{\leq} -: 0 \rightarrow 1$ and $\boxed{\geq} \circ -: 1 \rightarrow 0$, denoting the relations $\{(\bullet, x) \mid x \geq 0\} \subseteq \mathbb{k}^0 \times \mathbb{k}^1$ and $\{(x, \bullet) \mid x \geq 0\} \subseteq \mathbb{k}^1 \times \mathbb{k}^0$, are two different representations for the same object in traditional algebra: the polyhedral cone $\{x \in \mathbb{k} \mid x \geq 0\} \subseteq \mathbb{k}^1$. Interestingly enough, applying the polar operator to them yields two different results. Analogous considerations hold for $\{x \in \mathbb{k} \mid x \leq 0\} \subseteq \mathbb{k}^1$.

$$(\boxed{\geq} \circ)^\circ = \boxed{\geq}^\circ; \circ^\circ = \text{diagram of } \boxed{\geq} \text{ with inputs and outputs swapped}; \bullet \text{---} \stackrel{\text{counl}}{=} \boxed{\geq} \circ \quad (23)$$

$$(\circ \boxed{\leq})^\circ = (\boxed{\geq} \circ)^{op} \stackrel{\text{Lemma 15}}{=} (\boxed{\geq} \circ)^\circ; \blacksquare \stackrel{(23)}{=} \boxed{\geq} \circ^{op}; \blacksquare \stackrel{\text{P6 zer}}{=} \circ \boxed{\geq} \quad (24)$$

$$(\circ \boxed{\geq})^\circ \stackrel{(24)}{=} (\circ \boxed{\leq})^\circ \stackrel{\text{Prop. 14.2}}{=} \circ \boxed{\leq} \quad (25)$$

$$(\circ \boxed{\leq} \circ)^\circ = (\circ \boxed{\geq} \circ)^{op} \stackrel{\text{Lemma 15}}{=} \blacksquare; (\circ \boxed{\geq} \circ)^\circ \stackrel{(25)}{=} \blacksquare; \circ \boxed{\leq} \circ^{op} \stackrel{\text{P6 zer}}{=} \circ \boxed{\leq} \circ \quad (26)$$

Observe that for the two diagrams above of type $1 \rightarrow 0$, \cdot° act as identity, while for those of type $0 \rightarrow 1$, it reverses the sign. This behaviour is justified by Proposition 11.

There is a number of other observations about the polar operator, which may be proven with graphical reasoning taking advantage of the inductive definition, the complete axiomatisation and the laws illustrated so far. While this material is not essential to our developments, we conclude this section with two simple “exercises” of that kind, which are left to the interested reader.

- **(Exercise 1)** Prove that, for all c in the form of (8), there exists some d in the form of (9) such that $c^\circ \stackrel{\text{PA}}{=} d$. *Hint:* use Lemma 13 and $cc-1$
- **(Exercise 2)** Prove that, for all c in the form of (10), there exists some d in the form of (11) such that $c^\circ \stackrel{\text{PA}}{=} d$. *Hint:* use (23).

4 Lemma of the alternatives

This section is devoted to the diagrammatic formulation of a lemma of *alternatives*, asserting that exactly one of two systems of linear inequalities (i.e. polyhedra) has a solution.

To approach the lemma, an important question is how to model “does a system have a solution?” in our graphical calculus. We focus attention on two morphisms of PDiag of type $0 \rightarrow 0$: the empty diagram \square and the diagram $\circ \dashv$. Intuitively, $\circ \dashv$ asserts that “ $0 = 1$ ”; its denotational semantics is the composition of the relations $\{(\bullet, 0)\}$ and $\{(1, \bullet)\}$, which gives the empty relation \emptyset . Since for any relation R in Rel_k , $R \oplus \emptyset = \emptyset = \emptyset \oplus R$, the behaviour of $\circ \dashv$ resembles that of a logical false. From the viewpoint of the equational theory, $\circ \dashv$ introduces an inconsistency; in particular, by means of the axiom \emptyset we are able to prove that $\circ \dashv \oplus c \stackrel{\text{PA}}{=} \circ \dashv \oplus d$ for any $c, d: n \rightarrow m$ in PDiag . As an example, consider the following equation:

$$\dashv \oplus \square \circ \dashv = \dashv \oplus \square \circ \dashv \stackrel{\emptyset}{=} \bullet \bullet \dashv \oplus \square \stackrel{P3}{=} \bullet \bullet \dashv \circ \dashv \stackrel{\circ \bullet - bo}{=} \bullet \circ \dashv \quad (27)$$

In an analogous way, the behaviour of the diagram \square can be regarded as a logical true. In particular, its semantics is the relation $id_0 = \{(\bullet, \bullet)\}$ which for any R in Rel_k is such that $R \oplus id_0 = R = id_0 \oplus R$.

Finally, note that in Rel_k the only possible morphisms of type $0 \rightarrow 0$ are exactly \emptyset and id_0 . Thus the following lemma holds.

► **Lemma 17** (From [5]). *For any diagram $c: 0 \rightarrow 0$ of PDiag , either $c \stackrel{\text{PA}}{=} \square$ or $c \stackrel{\text{PA}}{=} \circ \dashv$*

► **Lemma 18** (Lemma of the alternatives). *Let $c: 0 \rightarrow 1$ be a diagram in PCDiag . Then exactly one of the following two equations holds:*

$$(a) \quad c; \dashv \stackrel{\text{PA}}{=} \square \quad (b) \quad c^\circ; \dashv \stackrel{\text{PA}}{=} \square.$$

Proof. Since c is in PCDiag , then $\llbracket c \rrbracket \subseteq k^0 \times k^1$ is a polyhedral cone. Thus $\llbracket c \rrbracket$ must be one of the following:

$$\{(\bullet, k) \mid k \in k\} \quad \{(\bullet, k) \mid k \geq 0\} \quad \{(\bullet, k) \mid k \leq 0\} \quad \{(\bullet, k) \mid k = 0\}$$

By Theorem 8, it holds¹ that either

$$c \stackrel{\text{PA}}{=} \bullet \dashv \quad \text{or} \quad c \stackrel{\text{PA}}{=} \circ \dashv \leq \dashv \quad \text{or} \quad c \stackrel{\text{PA}}{=} \circ \dashv \geq \dashv \quad \text{or} \quad c \stackrel{\text{PA}}{=} \circ \dashv.$$

¹ See Appendix A for a purely equational proof that does not invoke completeness.

By Proposition 14.1, we can thus consider only these four cases:

- If $c \stackrel{\mathbb{P}\mathbb{A}}{=} \bullet$ then $c; \dashv \stackrel{del}{=} \square$ and $c^\circ; \dashv \stackrel{Prop. 14}{=} (\bullet)^\circ; \dashv = \circ$.
- If $c \stackrel{\mathbb{P}\mathbb{A}}{=} \circ \text{---} \langle \leq \text{---} \rangle$ then $c; \dashv \stackrel{AP1}{=} \square$ and $c^\circ; \dashv \stackrel{Prop. 14}{=} (\circ \text{---} \langle \leq \text{---} \rangle)^\circ; \dashv \stackrel{(24)}{=} \circ \text{---} \langle \geq \text{---} \rangle; \dashv \stackrel{(14)}{=} \circ$.
- If $c \stackrel{\mathbb{P}\mathbb{A}}{=} \circ \text{---} \langle \geq \text{---} \rangle$ then $c; \dashv \stackrel{(14)}{=} \circ$ and $c^\circ; \dashv \stackrel{Prop. 14}{=} (\circ \text{---} \langle \geq \text{---} \rangle)^\circ; \dashv \stackrel{(25)}{=} \circ \text{---} \langle \leq \text{---} \rangle; \dashv \stackrel{AP1}{=} \square$.
- If $c \stackrel{\mathbb{P}\mathbb{A}}{=} \circ$ then $c; \dashv \stackrel{\mathbb{P}\mathbb{A}}{=} \circ$ and $c^\circ; \dashv \stackrel{Prop. 14}{=} (\circ)^\circ; \dashv = \bullet; \dashv \stackrel{del}{=} \square$. ◀

The lemma of alternatives yields as a corollary a proof of Proposition 11.

Proof of Proposition 11. Observe that

$$\boxed{c^\circ} \text{---} \boxed{\leftarrow} \text{---} \boxed{b} \text{---} \dashv \stackrel{\mathbb{P}\mathbb{A}}{=} \square \stackrel{Lemma 18}{\iff} \left(\boxed{c^\circ} \text{---} \boxed{\leftarrow} \text{---} \boxed{b} \right)^\circ; \dashv \stackrel{\mathbb{P}\mathbb{A}}{=} \circ \dashv \stackrel{Lemma 13}{\iff} \boxed{c} \text{---} \boxed{\rightarrow} \text{---} \boxed{b^T} \text{---} \dashv \stackrel{\mathbb{P}\mathbb{A}}{=} \circ \dashv$$

By definition of $\llbracket \cdot \rrbracket$, the former equation holds iff $(\bullet, b) \in \llbracket c^\circ \rrbracket$, while the latter holds iff $\forall (\bullet, x) \in \llbracket c \rrbracket, b^T x \neq 1$. That is $\llbracket c^\circ \rrbracket$ is the relation $\{(\bullet, b) \mid \forall x \in C, b^T x \neq 1\}$ which is readily seen to be equal to $\{(\bullet, b) \mid b \in polar(C)\}$.

For d , note that $d \stackrel{\mathbb{P}\mathbb{A}}{=} c^{op}$. Thus, by Lemma 15, $d^\circ \stackrel{\mathbb{P}\mathbb{A}}{=} n \text{---} \blacksquare \text{---} n; (c^\circ)^{op}$. Thus $(b, \bullet) \in \llbracket d^\circ \rrbracket$ iff $(\bullet, -b) \in \llbracket c^\circ \rrbracket$ iff $-b \in polar(C)$ iff $b \in dual(C)$. That is $\llbracket d^\circ \rrbracket = \{(b, \bullet) \mid b \in dual(C)\}$. ◀

► **Remark 19.** Interestingly, the lemma of alternatives does not hold for diagrams $c: 1 \rightarrow 0$ (when taking $\vdash; c$ and $\vdash; c^\circ$ in place of $c; \dashv$ and $c^\circ; \dashv$): it is easy to see this with (23) and (26). In order to obtain a lemma of alternatives for diagrams of type $c: 1 \rightarrow 0$, one should replace \cdot° by a novel operator \cdot^\bullet defined as $\text{---} \langle \geq \text{---} \rangle^\bullet = \text{---} \langle \leftarrow \text{---} \rangle; (\text{---} \langle \leq \text{---} \rangle \oplus \text{---})$ and as $c^\bullet = c^\circ$ for all the other generators c . Such operator behaves as the dual for diagrams $c: 0 \rightarrow n$ and as the polar for diagrams $d: n \rightarrow 0$.

5 A string diagrammatic proof of Farkas' Lemma

The lemma of alternatives provides a direct route to a diagrammatic proof of Farkas' lemma.

► **Lemma 20 (Farkas' lemma).** Let $\vec{A}: n \rightarrow m$ be a diagram in $MDiag$ and $\overleftarrow{b}: m \rightarrow 1$ in $MDiag$, then exactly one of the following two equations holds:

$$(a) \quad \circ \text{---} \langle \leq \text{---} \rangle \text{---} \boxed{\vec{A}} \text{---} \boxed{\overleftarrow{b}} \text{---} \dashv \stackrel{\mathbb{P}\mathbb{A}}{=} \square \quad (b) \quad \circ \text{---} \langle \leq \text{---} \rangle \text{---} \boxed{\overleftarrow{A^T}} \text{---} \boxed{\vec{b^T}} \text{---} \blacksquare \dashv \stackrel{\mathbb{P}\mathbb{A}}{=} \square$$

Proof. Observe that $\circ \text{---} \langle \leq \text{---} \rangle \text{---} \boxed{\vec{A}} \text{---} \boxed{\overleftarrow{b}} \text{---} \dashv$ is a diagram $c: 0 \rightarrow 1$ in $PCDiag$. In order to conclude, it is therefore enough to use Lemma 18 and observe that

$$\begin{aligned} \left(\circ \text{---} \langle \leq \text{---} \rangle \text{---} \boxed{\vec{A}} \text{---} \boxed{\overleftarrow{b}} \text{---} \dashv \right)^\circ &= \circ \text{---} \langle \geq \text{---} \rangle \text{---} \boxed{\overleftarrow{A^T}} \text{---} \boxed{\vec{b^T}} \text{---} \dashv && \text{(Lemma 13 and (24))} \\ &\stackrel{\mathbb{P}\mathbb{A}}{=} \circ \text{---} \langle \geq \text{---} \rangle \text{---} \boxed{\overleftarrow{A^T}} \text{---} \boxed{\vec{b^T}} \text{---} \blacksquare \blacksquare \text{---} && \text{((17))} \\ &\stackrel{\mathbb{P}\mathbb{A}}{=} \circ \text{---} \langle \geq \text{---} \rangle \text{---} \blacksquare \text{---} \boxed{\overleftarrow{A^T}} \text{---} \boxed{\vec{b^T}} \text{---} \blacksquare \text{---} && \text{((16))} \\ &\stackrel{\mathbb{P}\mathbb{A}}{=} \circ \text{---} \langle \leq \text{---} \rangle \text{---} \boxed{\overleftarrow{A^T}} \text{---} \boxed{\vec{b^T}} \text{---} \blacksquare \text{---} && \text{(Axioms P6 and del)} \end{aligned}$$

◀

It is instructive to make explicit in which sense Lemma 20 amounts to the well known result of Farkas. By using the inductive definition of $\llbracket \cdot \rrbracket$, one may compute the semantics on the left hand sides of the equations (a) and (b):

$$\begin{aligned} \llbracket \text{circ} \text{---} \text{box}(\leq) \text{---} \text{box}(\overrightarrow{A}) \text{---} \text{box}(\overleftarrow{b}) \text{---} \text{box}(\dashv) \rrbracket &= \begin{cases} \{(\bullet, \bullet)\} & \text{if } \exists x \in \mathbb{k}^n \text{ s.t. } x \geq 0 \text{ and } Ax = b \\ \emptyset & \text{otherwise} \end{cases} \\ \llbracket \text{circ} \text{---} \text{box}(\leq) \text{---} \text{box}(\overleftarrow{A^T}) \text{---} \text{box}(\overrightarrow{b^T}) \text{---} \text{box}(\blacksquare) \text{---} \text{box}(\dashv) \rrbracket &= \begin{cases} \{(\bullet, \bullet)\} & \text{if } \exists y \in \mathbb{k}^m \text{ s.t. } A^T y \geq 0 \text{ and } b^T y = -1 \\ \emptyset & \text{otherwise} \end{cases} \end{aligned}$$

Therefore equation (a) holds if and only if $\exists x \in \mathbb{k}^n$ s.t. $x \geq 0$ and $Ax = b$ while equation (b) if and only if $\exists y \in \mathbb{k}^m$ s.t. $A^T y \geq 0$ and $b^T y = -1$. In the usual presentation of the Farkas' lemma, e.g. [20], the former condition is exactly the same, while the second one is often expressed by the equivalent condition $\exists y \in \mathbb{k}^m$ s.t. $A^T y \geq 0$ and $b^T y < 0$.²

6 Duality in linear programming

Farkas' lemma is closely related to *linear programming*, as it is one of the main tools to prove duality results in this area. In this section, we explore such duality theorems in the context of diagrammatic polyhedral algebra; it turns out that our formulation does not require the direct application of Farkas' lemma, but rather relies on more general principle (Theorem 23) which allows to prove all the results at once.

Duality in linear programming studies pairs of problems of the following shape

$$(P) := \max\{cx \mid Ax \leq b, x \geq 0\} \qquad (D) := \min\{b^T y \mid A^T y \geq c^T \text{ and } y \geq 0\}$$

where A , b and c are matrices of type $m \times n$, $1 \times m$ and $n \times 1$, respectively. The *primal problem* (P) requires to maximise cx subject to the condition that $Ax \leq b$ and $x \geq 0$. Its *dual problem* (D) requires to minimise $b^T y$ subject to the condition that $A^T y \geq c^T$ and $y \geq 0$. The former problem and the one of the board from the Introduction are instances of (P) and (D) . The primal problem has three possible outcomes: (P) may be *unfeasible*, in the sense that there exists no $x \geq 0$ such that $Ax \leq b$; it can be *unbounded*, when the latter inequality holds for some non-negative vectors $x \in \mathbb{k}^n$, but there exists no maximum $k \in \mathbb{k}$ for cx ; or it can be *bounded*, if such k exists. The same possibilities apply to (D) .

Duality theory in linear programming establishes a series of possibilities between these possible outcomes: in particular, if (P) is unbounded then (D) is unfeasible and, viceversa, if (D) is unbounded then (P) is unfeasible. Moreover, (P) is bounded if and only if (D) is bounded. The following table summarises such results.

(P)	(D)	bounded	unbounded	unfeasible
bounded		✓		
unbounded				✓
unfeasible			✓	✓

(28)

² By mean of Proposition 11 one can also translate our proof in traditional algebraic language: first observe that for all one-dimensional polyhedral cones C either 1 belongs to C or 1 belong to the polar of C (Lemma 18); then prove that the polar of $\{z \in \mathbb{k} \mid \exists x \in \mathbb{k}^n \text{ s.t. } x \geq 0 \text{ and } Ax = bz\}$ is exactly $\{z \in \mathbb{k} \mid \exists y \in \mathbb{k}^m \text{ s.t. } A^T y \geq 0 \text{ and } b^T y = -z\}$ (proof of Lemma 20). We could not find the same proof in literature, but it is hard to claim that it does not exist.

9:14 From Farkas' Lemma to Linear Programming: An Exercise in Diagrammatic Algebra

The most useful fact is that when (P) and (D) are bounded, they have the same result, i.e.,

$$\max\{cx \mid Ax \leq b, x \geq 0\} = \min\{b^T y \mid A^T y \geq c^T \text{ and } y \geq 0\}. \quad (29)$$

We now turn to the question of modelling the primal problem (P) and the dual problem (D) in PDiag. Let us fix $\overleftarrow{A}: m \rightarrow n$ in MDiag $\overleftarrow{}$, $\overrightarrow{b}: 1 \rightarrow m$ and $\overleftarrow{c}: n \rightarrow 1$ in MDiag $\overrightarrow{}$, and consider the following diagrams in PDiag



Their semantics can be easily computed with the inductive definition in (6):

$$\begin{aligned} \llbracket P \rrbracket &= \{(\bullet, z) \in \mathbf{k}^0 \times \mathbf{k}^1 \mid z \leq cx, x \geq 0, Ax \leq b\} \\ \llbracket D \rrbracket &= \{(z, \bullet) \in \mathbf{k}^1 \times \mathbf{k}^0 \mid z \geq b^T y, y \geq 0, A^T y \geq c^T\} \end{aligned}$$

As expected, P models the primal problem (P) and D its dual (D) . Indeed, (P) is bounded if and only if $\llbracket P \rrbracket = \{(\bullet, z) \mid z \leq k\}$, where k is exactly $\max\{cx \mid Ax \leq b, x \geq 0\}$. Also, (P) is unbounded if and only if $\llbracket P \rrbracket = \{(\bullet, z) \mid z \in \mathbf{k}\}$ and (P) is unfeasible if and only if $\llbracket P \rrbracket = \emptyset$. Analogous considerations hold for (D) . The three possibilities can then be expressed in equational terms as follows.

$$\begin{array}{l} P \stackrel{\text{PA}}{=} \vdash \boxed{k} \boxed{\geq} \dashv \text{ iff } k = \max\{cx \mid Ax \leq b, x \geq 0\} \\ P \stackrel{\text{PA}}{=} \bullet \dashv \text{ iff } (P) \text{ is unbounded} \\ P \stackrel{\text{PA}}{=} \vdash \circ \dashv \text{ iff } (P) \text{ is unfeasible} \end{array} \quad \left| \quad \begin{array}{l} D \stackrel{\text{PA}}{=} \dashv \boxed{\geq} \boxed{k} \vdash \text{ iff } k = \min\{b^T y \mid A^T y \geq c^T, y \geq 0\} \\ D \stackrel{\text{PA}}{=} \bullet \dashv \text{ iff } (D) \text{ is unbounded} \\ D \stackrel{\text{PA}}{=} \bullet \circ \dashv \text{ iff } (D) \text{ is unfeasible} \end{array} \right.$$

In light of this analysis, the results in (28) and (29) amount to the following theorem.

► **Theorem 21 (Duality).** *The following hold:*

1. For all $k \in \mathbf{k}$, $P \stackrel{\text{PA}}{=} \vdash \boxed{k} \boxed{\geq} \dashv$ if and only if $D \stackrel{\text{PA}}{=} \dashv \boxed{\geq} \boxed{k} \vdash$
2. If $P \stackrel{\text{PA}}{=} \bullet \dashv$, then $D \stackrel{\text{PA}}{=} \bullet \circ \dashv$
3. If $D \stackrel{\text{PA}}{=} \bullet \dashv$, then $P \stackrel{\text{PA}}{=} \vdash \circ \dashv$

In order to prove the above theorem, we exploit homogenisation, a traditional technique to transform polyhedra into cones. The *homogenisation* of polyhedron $P = \{x \in \mathbf{k}^n \mid Ax + b \geq 0\}$ is the polyhedral cone $P^H = \{(x, y) \in \mathbf{k}^{n+1} \mid Ax + by \geq 0, y \geq 0\}$. It holds that $P_1^H = P_2^H$ if and only if $P_1 = P_2$ for all *non-empty* polyhedra P_1, P_2 (see e.g. Lemma 22 in [5]). By exploiting the normal forms in (12) and (10), one obtains the following useful lemma.

► **Lemma 22.** *Let $c, d: n + 1 \rightarrow m$ be string diagrams in PCDiag.*

1. If $\vdash \boxed{c} \boxed{\geq} \dashv \stackrel{\text{PA}}{=} \vdash \boxed{d} \boxed{\geq} \dashv$ then $\vdash \boxed{c} \boxed{\geq} \dashv \stackrel{\text{PA}}{=} \vdash \boxed{d} \boxed{\geq} \dashv$;
2. Moreover when $\llbracket \vdash \boxed{c} \boxed{\geq} \dashv \rrbracket \neq \emptyset$, the other implication holds, that is:

$$\vdash \boxed{c} \boxed{\geq} \dashv \stackrel{\text{PA}}{=} \vdash \boxed{d} \boxed{\geq} \dashv \text{ iff } \vdash \boxed{c} \boxed{\geq} \dashv \stackrel{\text{PA}}{=} \vdash \boxed{d} \boxed{\geq} \dashv.$$

By combining homogenisation with the polar operator, we obtain a general proof schema which includes as a particular cases the three points of Theorem 21.

Thanks to Theorem 23, we do not really need to rely on a specific form. Indeed by taking

$$P' := \vdash \boxed{\vec{b}} \text{---} \boxed{\geq} \text{---} \boxed{\overleftarrow{A}} \text{---} \boxed{\vec{c}} \text{---} \boxed{\geq} \text{---} \dashv$$

$$D' := \dashv \boxed{\geq} \text{---} \boxed{\overleftarrow{b^T}} \text{---} \boxed{\overleftarrow{A^T}} \text{---} \boxed{\overleftarrow{c^T}} \text{---} \dashv$$

one can easily check that Theorem 21 holds also for P' and D' and the proof is the same, modulo the obvious choice of c .

7 Conclusions

This paper investigates Farkas' lemma and duality in linear programming within the language of diagrammatic polyhedral algebra. Besides the elegance of the proofs, the linguistic aspect is, in our opinion, the most interesting angle. Indeed, this work can be thought as an exercise in diagrammatic algebra, illustrating its appeal in the following ways:

- by identifying the right primitive components and the appropriate ways to compose them, one is able to express exactly all the objects of interest (in this case, polyhedra) and to formally reason about them by means of a sound and complete axiomatisation ($\mathbb{P}\mathbb{A}$);
- operations on few primitives can be extended inductively to all the objects of interest, resulting in an effective way to compute sophisticated notions, like polar and dual cones;
- equations amongst diagrams can express complex statements, like those about the existence of a solution, the maximal or the minimal solution;
- symbolic manipulation of diagrams by means of axioms and derived laws allows to prove such statements.

The last point leads us to believe that our proofs are suitable to be formalised in proof assistants, such as Coq or Agda. Finally, we think that this work may inspire further duality results in string diagrammatic languages other than diagrammatic polyhedral algebra.

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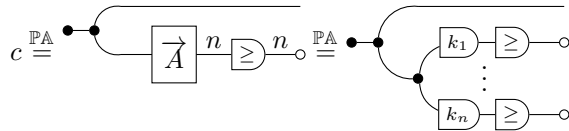
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A Proofs of Section 4

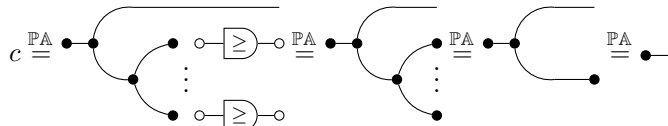
Alternative proof of Lemma 18. We prove diagrammatically, without relying on Theorem 8, that the following property holds for any $c: 0 \rightarrow 1$ in PCDiag

$$c \stackrel{\mathbb{P}\mathbb{A}}{=} \bullet \text{---} \quad \text{or} \quad c \stackrel{\mathbb{P}\mathbb{A}}{=} \circ \text{---} \langle \leq \text{---} \quad \text{or} \quad c \stackrel{\mathbb{P}\mathbb{A}}{=} \circ \text{---} \langle \geq \text{---} \quad \text{or} \quad c \stackrel{\mathbb{P}\mathbb{A}}{=} \circ \text{---}.$$

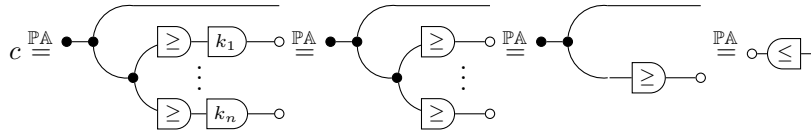
Any $n \rightarrow m$ diagram in PCDiag is equivalent to one in the form of (10). A diagram $c: 0 \rightarrow 1$ has the following normal form, where A is a $n \times 1$ matrix



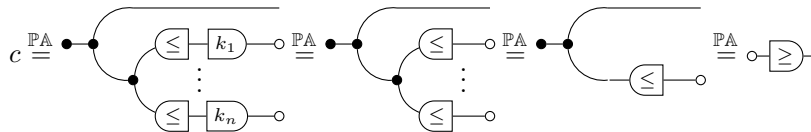
- If $k_i = 0$ for all $i = 1 \dots n$, then



- If $k_i \geq 0$ for all $i = 1 \dots n$, then those $k_i = 0$ are just *detached* as in the first case and for the others



- If $k_i \leq 0$ for all $i = 1 \dots n$, then those $k_i = 0$ are just *detached* as in the first case and for the others



- If some $k_i \geq 0$, and some $k_i \leq 0$, then those $k_i = 0$ are just *detached* as in the first case. For the others, assume without loss of generality that the first j are positive and the remaining ones are negative.

