# Chipsplitting Games: A Combinatorial Approach to Classifying One-Dimensional Discrete Statistical Models with Rational Maximum Likelihood Estimator

by

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## Zusammenfassung in deutscher Sprache

Die vorliegende Ausarbeitung setzt die Forschung von Arthur Bik und Orlando Marigliano zur Klassifizierung ein-dimensionaler diskreter statistischer Modelle mit rationalen Maximum Likelihood Schätzern unter Verwendung fundamentaler Modelle fort. Wir erzielen bedeutende Fortschritte beim Beweis zur endlichen Anzahl der fundamentalen Modelle im Wahrscheinlichkeitssimplex  $\Delta_5$ . Zudem bestimmen wir die Anzahl der fundamentalen Modelle im Simplex  $\Delta_6$  mit einem maximalen Grad von 11.

#### Abstract

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This paper continues the research of Arthur Bik and Orlando Marigliano on the classification of one-dimensional discrete statistical models with rational maximum likelihood estimators using fundamental models. We present a missing proof of an algorithm from their work. Furthermore, we make significant progress in proving the finite number of fundamental models in the probability simplex  $\Delta_5$ . We also determine the number of fundamental models in the simplex  $\Delta_6$  with a maximum degree of 11.

# Contents

Co	Contents						
1	Introduction						
2	Classification with Fundamental Models2.1 Parametrization2.2 Reduced Models2.3 Fundamental Models2.4 On the Finiteness of Fundamental Models	3 3 7 8 11					
<b>3 4</b>	Chipsplitting Games  3.1 Basic Definitions	14 14 17 22 29					
	4.2 Divide and Conquer4.3 Implementation of the Hyperfield Criterion4.4 Symmetry4.5 Impossible Supports	31 34 34 36					
5	Valid Outcomes of Positive Support Size One, Two, and Three	40					
6	Constraints on Supports of Valid Outcomes 6.1 Hyperfield Criterion	44 44 50 54 57					
7	Valid Outcomes of Positive Support Size Four	66					
8	Hexagon Criterion	69					

9	Valid Outcomes of Positive Support Size Five	72
	9.1 Case $d = 8,, 41$	72
	9.2 Case $d \ge 42$	73
	9.3 Case $d \ge 42$ Continued: $\mathbf{s}' \in \Lambda$ with Positive Support Size Four	. 76
	9.4 Case $d \ge 42$ Continued: $\mathbf{s}' \in \Lambda$ with Positive Support Size Five	76
10	O Valid Outcomes of Positive Support Size Six	91
	10.1 Fixed-Contractable Pascal Forms	93
	10.2 An Extended Trivial System	. 99
	10.3 Reducing Cases with Fixed-Contractable Pascal Forms	100
11	1 Discussion	101
Bi	ibliography	103

# Chapter 1

## Introduction

In statistics we come across various collections of probability distributions, such as the normal distribution, Poisson distribution, and binomial distribution. These distributions are used to model random variables in applications, and are referred to as *statistical models*. Precisely, a statistical model is just a set of probability distributions. If the set contains only discrete distributions, we call it a *discrete statistical model*. In this case, discrete statistical models are just subsets of the probability simplex  $\Delta_n := \{p \in \mathbb{R}^{n+1} \mid \sum p_i = 1\}$ .

A discrete distribution  $p \in \mathcal{M} \subset \Delta_n$  from a discrete statistical model encapsulates the probabilities of observing the states  $0, \ldots, n$ , i.e. if  $X \in \{0, \ldots, n\}$  is a discrete random variable, then the state X = i occurs with probability  $p_i$  for all  $i = 0, \ldots, n$ . Say we have a binomial random variable X with n + 1 states, then  $p_i = \binom{n}{i} \theta^i (1 - \theta)^{n-i}$  computes the probability of observing i successes in n trials with success probability  $\theta \in [0, 1]$ . The set  $\mathcal{M}$  of all probability distributions of that form, i.e.  $\mathcal{M} = \{\binom{n}{i} \theta^i (1 - \theta)^{n-i}\}_{i=0}^n \mid \theta \in [0, 1]\}$ , is our first example of a discrete statistical model, and is known as the binomial model.



Figure 1.1: This figure shows the probability simplex  $\Delta_2$  with the binomial model (red curve). Every point on the curve is a binomial distribution.

Given a statistical model  $\mathcal{M} \subset \Delta_n$  and data  $u \in \mathbb{N}^{n+1}$ , a typical problem in statistics is to find a distribution from a statistical model that best describes the data. "Best" can mean a lot of things, but in maximum likelihood estimation it means finding the distribution that maximizes the probability of observing the data; the map  $\Phi : \Delta_n \to \mathcal{M}, u \mapsto \hat{p}$  that assigns the data u to a distribution  $\hat{p} \in \mathcal{M}$  from the statistical model is called the maximum likelihood estimator (MLE). This map is characterized by the property that  $\hat{p}$  maximizes the log-likelihood function  $\ell(p) = \sum u_i \log p_i$  for all  $p \in \mathcal{M}$ .

We focus on one-dimensional discrete statistical models with rational MLE. These are models  $\mathcal{M}$  satisfying

- $\mathcal{M} = \text{image}(p)$  for some rational map  $p = (p_0, \dots, p_n) : I \to \Delta_n$  where  $p_i$  is rational,  $I \subset \mathbb{R}$  is a union of closed intervals and  $p(\partial I) \subset \partial \Delta_n$ ,
- all the n+1 coordinates of the maximum likelihood estimator  $\Phi$  are rational functions in the data u.

There are two intriguing questions to ask about statistical models with rational MLE: the first one is about which form do the maximum likelihood estimators take; the second one is more concerned with the classification of the statistical models, i.e. can we divide these models into easier to understand classes? An answer to the first question was given by June Huh. He showed that if  $\Phi$  is rational, then each of its coordinates is an alternating product of linear forms with numerator and denominator of the same degree, see [5, 3]. For the second question, Arthur Bik and Orlando Marigliano classified all one-dimensional discrete statistical models with rational MLE using fundamental models [2].

This thesis continues the work of Bik and Marigliano. In the first half, we present their classification results on how fundamental models serve as the building blocks of one-dimensional discrete models with rational MLE. In the second half, we establish and extend their finding that there are only finitely many fundamental models within the probability simplices  $\Delta_n$  for  $n \leq 4$ . Due to the complexity of the problem, the cases  $n \geq 5$  were left open. For the first time, we make significant progress for n = 5. Other contributions of this thesis include a new result on the number of fundamental models in  $\Delta_6$  with a maximum degree of 11, and a missing proof of a key algorithm from their work.

## Chapter 2

# Classification with Fundamental Models

In this chapter we present the classification of one-dimensional discrete statistical models with rational maximum likelihood estimator (MLE) using fundamental models. The classification is due to Arthur Bik and Orlando Marigliano [2].

**Problem statement:** Can we find a class of easy to understand models that serve as building blocks for all one-dimensional discrete statistical models with rational MLE?

The answer to this question are reduced and fundamental models.

#### 2.1 Parametrization

It turns out that one-dimensional discrete statistical models with rational MLE admit the following parametrization.

**Proposition 2.1.** Let  $\mathcal{M}$  be a one-dimensional discrete statistical models with rational maximum likelihood estimator. Then, there exists a map of the form

$$p: [0,1] \to \Delta_n, \quad \theta \mapsto (w_k \theta^{i_k} (1-\theta)^{j_k})_{k=0}^n$$
  
 $i_k, j_k \in \mathbb{Z}_{>0}, \ w_k \in \mathbb{R}_{>0} \quad \forall k = 0, \dots, n$ 

such that  $\mathcal{M} = \text{image}(p)$ .

We introduce some notation to simplify the proof of Proposition 2.1. Let  $\mathcal{M} \subset \Delta_n$  be a one-dimensional discrete statistical model parametrized by rational functions  $p_0 = \frac{g_0}{h_0}, \ldots, p_n = \frac{g_n}{h_n}$ . Define b to be the least common multiple of  $h_0, \ldots, h_n$  and  $a_i \coloneqq bp_i$ . Since  $\sum p_k = 1$ , we can multiply by b to obtain  $\sum a_k = b$ . We see that the polynomials  $a_0, \ldots, a_n, b$ 

determine the statistical model  $\mathcal{M}$ , and have no common factors. The log-likelihood function is then given by

$$\ell(p) = \sum u_i \log p_i$$

$$= \sum u_i \log \frac{a_i}{b}$$

$$= \sum u_i \log a_i - \sum u_i \log b.$$

To find the maximum likelihood estimator, we need find all critical points of the log-likelihood function. This is equivalent to finding the roots of the gradient of the log-likelihood function

$$\ell(p(\theta))' = \sum u_k \frac{a_k'}{a_k} - \sum u_k \frac{b'}{b} = 0.$$
 (2.1)

These equations are called the *score equations* in algebraic statistics, and the number of complex solutions to these equations for general data  $u \in \mathbb{C}^{n+1}$  is called the *maximum likelihood degree* of the statistical model. This ML degree has an important meaning in algebraic statistics, as it determines the complexity of the model. We have the following relationship between the ML estimator and the ML degree.

**Proposition 2.2.** Having rational maximum likelihood estimator can be expressed equivalently by saying that the maximum likelihood degree of the statistical model is one.

To prove Proposition 2.1, we need the following lemma.

**Lemma 2.3.** If  $\mathcal{M}$  has rational MLE, then there are exactly two distinct complex linear factors in  $a_0, \ldots, a_n$ , and b.

*Proof.* We prove the lemma in three steps:

• Let f be the product of all distinct complex linear factors in  $a_0, \ldots, a_n, b$ . If we multiply the score equations (2.1) by f, we get

$$f \cdot \ell(p(\theta))' = \sum u_k f \frac{a'_k}{a_k} - \sum u_k f \frac{b'}{b} = 0.$$

Note that every linear factor of  $a_k$  with multiplicity m occurs in  $a'_k$  with multiplicity m-1; thus every summand of  $\frac{a'_k}{a_k}$  is of the form  $\frac{\lambda}{(x-\xi)}$ , where  $\lambda \in \mathbb{R}$  and  $x-\xi$  is some linear factor of  $a_k$ ; hence  $f \cdot \frac{\lambda}{(x-\xi)}$  is of degree  $\deg(f)-1$ , and therefore  $f \cdot \ell(p(\theta))'$  is of degree  $\deg(f)-1$ .

- We claim that the roots of  $\ell(p(\theta))'$  are the same as the roots of  $f \cdot \ell(p(\theta))'$ . Assume we have shown this claim. By Proposition 2.2 the ML degree is one. So,  $\ell(p(\theta))'$  has one root. Thus,  $f \cdot \ell(p(\theta))'$  has one root, and therefore  $f \cdot \ell(p(\theta))'$  is of degree one. This implies that  $\deg(f) = 2$  with the previous step. Thus, there are exactly two distinct complex linear factors in  $a_0, \ldots, a_n$ , and b.
- It remains to show that the roots stay the same. Clearly, every root of  $\ell(p(\theta))'$  is a root of  $f \cdot \ell(p(\theta))'$ . Conversely, we want to show that no new roots are introduced when multiplying by f, i.e. roots of f are not roots of  $f \cdot \ell(p(\theta))'$ . To do so, we rewrite

$$f \cdot \ell(p(\theta))' = \sum_{k=0}^{n} u_k f \frac{a'_k}{a_k} - \sum_{k=0}^{n} u_k f \frac{b'}{b} = \sum_{k=0}^{n+1} v_k f \frac{c'_k}{c_k}$$
$$v_k = u_k, \ c_k = a_k \quad \text{for } k = 0, \dots, n,$$
$$v_{n+1} = -\sum_{k=0}^{n} u_k, \ c_{n+1} = b.$$

Let q be a complex linear factor of f. We define polynomials  $r_0, \ldots, r_{n+1}$  and r such that  $c_k = q^{l_k} r_k$ , f = qr, and  $r_0, \ldots, r_{n+1}, r$  do not have q as a factor. Then, we have for  $k = 0, \ldots, n+1$  that

$$f\frac{c'_k}{c_k} = qr \cdot \frac{l_k q^{l_k - 1} q' r_k + q^{l_k} r'_k}{q^{l_k} r_k} = qr \frac{l_k q'}{q} + qr \frac{r'_k}{r_k} \equiv r l_k q' \pmod{q}.$$

Thus, we obtain

$$f \cdot \ell(p(\theta))' \equiv rq' \sum_{k=0}^{n+1} v_k l_k \equiv rq' \sum_{k=0}^{n} v_k (l_k - l_{n+1}) \pmod{q}.$$

Note that by definition of  $l_k$ , a value of  $l_k = 0$  means that q is not a factor of  $c_k$ . By definition of f, at least one  $l_k > 0$ . On the other hand, not all  $l_k$  can be positive since  $a_0, \ldots, a_n, b$  share no common factors. Hence, not all  $l_k - l_{n+1} = 0$  vanish. Hence, for generic data u we assume  $\sum_{k=0}^{n} v_k (l_k - l_{n+1}) \neq 0$ . This with  $q'r \not\equiv 0 \pmod{q}$  implies that q is not a complex linear factor of  $f \cdot \ell(p(\theta))'$ . We showed that the roots of f are not roots of  $f \cdot \ell(p(\theta))'$ .

Equipped with the lemma, we can now prove Proposition 2.1.

*Proof.* We want to show the following parametrization of  $\mathcal{M}$ :

$$p:[0,1]\to\Delta_n,\quad\theta\mapsto(w_k\theta^{i_k}(1-\theta)^{j_k})_{k=0}^n$$

First, we show that I is a single closed real interval and not a union of closed intervals. For the sake of contradiction assume that  $I = \bigcup_k I_k$  is a union of closed disjoint intervals. By definition of  $\mathcal{M}$  we know that  $p(\partial I) \subset \partial \Delta_n$ . Thus, there exist  $\theta_1, \theta_2 \in \partial I_0$  and  $\theta_3, \theta_4 \in \partial I_1$  with  $p_i(\theta_1) = p_i(\theta_2) = 0$  and  $p_j(\theta_3) = p_j(\theta_4) = 0$  for some  $i, j = 0, \ldots, n$ . Note that  $\theta_1, \theta_2$  are roots of  $\frac{a_i}{b}$  and  $\theta_3, \theta_4$  are roots of  $\frac{a_j}{b}$ . By Lemma 2.3 exactly two distinct complex linear factors occur in  $a_0, \ldots, a_n, b$ . Hence,  $\theta_3 = \theta_1$  or  $\theta_3 = \theta_2$ . Contradiction for  $I_0$  and  $I_1$  are disjoint.

The previous argument shows that  $I = [\alpha, \beta]$  is a real single closed interval. Thus, the roots of  $a_0, \ldots, a_n, b$  are real and take values in  $\partial I = \{\alpha, \beta\}$ . By a suitable parametrization, we can assume without loss of generality that I = [0, 1]. We can now write the polynomials  $a_0, \ldots, a_n, b$  as

$$a_k(\theta) = w_k \theta^{i_k} (1 - \theta)^{j_k}$$
$$b(\theta) = w \theta^i (1 - \theta)^j$$

with  $w_k, w \in \mathbb{R}_{>0}$ , and  $i_k, j_k, i, j \in \mathbb{Z}_{\geq 0}$  for all k = 0, ..., n. Since  $a_0, ..., a_n, b$  share no common factors, there exists some  $i_k = 0$  if i > 0; however this would contradict  $0 < w_k \le a_0(0) + \cdots + a_n(0) = b(0) = 0$ . So i = 0. Similarly, j = 0. Finally, we divide p by w to obtain  $b \equiv 1$ .

**Corollary 2.4.** Any one-dimensional discrete statistical models with rational MLE can be represented by  $(w_k, i_k, j_k)_{k=0}^n$  for  $w_k \in \mathbb{R}_{>0}$  and  $i_k, j_k \in \mathbb{Z}_{\geq 0}$ .

From now on, we only consider one-dimensional discrete statistical models with rational MLE; we call them *models* for short.

**Definition 2.5.** The degree  $\deg(\mathcal{M})$  of a model  $\mathcal{M}$  represented by  $(w_k, i_k, j_k)_{k=0}^n$  is defined as  $\max\{i_k + j_k : k = 0, \dots, n\}$ .

**Remark 2.6.** We view two models  $(w_k, i_k, j_k)_{k=0}^n$  and  $(w'_k, i'_k, j'_k)_{k=0}^n$  as the same model if they are equal up to a permutation of the coordinates.

**Example 2.7.** The sequence ((1,0,2),(2,1,1),(1,2,0)) represents the binomial model with two trials. It has degree two. Its parametrization is given by  $\theta \mapsto ((1-\theta)^2, 2\theta(1-\theta), \theta^2)$ . Also see Figure 1.1 for a visualization of the binomial model within the probability simplex  $\Delta_2$ .

Note that we view the sequences ((1,0,2),(2,1,1),(1,2,0)) or ((2,1,1),(1,0,2),(1,2,0)) as the same model as ((2,1,1),(1,2,0),(1,0,2)) since the order of the coordinates does not matter.

**Definition 2.8.** Let  $\mathcal{M}$  be a model represented by  $(w_k, i_k, j_k)_{k=0}^n$ . The set of exponent pairs  $(i_k, j_k)_{k=0}^n$  is called the support of  $\mathcal{M}$ , denoted by  $\sup(\mathcal{M})$ .

This was our first step towards understanding the structure of models. The next step is to introduce the concept of reduced models.

#### 2.2 Reduced Models

Models in this section refer to one-dimensional discrete statistical models with rational MLE.

**Definition 2.9.** We call a model represented by  $(w_k, i_k, j_k)_{k=0}^n$  reduced if  $(i_k, j_k) \neq \mathbf{0}$  for all  $k = 0, \ldots n$ , and  $(i_k, j_k) \neq (i_l, j_l)$  for all  $k \neq l$ .

Due to  $(i_k, j_k) \neq (i_l, j_l)$ , we can use functions to represent reduced models.

**Remark 2.10.** A reduced model  $\mathcal{M}$  represented by  $(w_k, i_k, j_k)_{k=0}^n$  can also be identified by a function  $f: \mathbb{Z}^2 \to \mathbb{R}_{\geq 0}, (i, j) \mapsto w$ , where  $w = w_k$  if  $(i_k, j_k) = (i, j)$  and w = 0 otherwise. The support of f is the set of all pairs (i, j) with f(i, j) > 0. It coincides with the support of  $\mathcal{M}$ .

Reduced models are our first building blocks for the classification of models. This statement is justified by the following two propositions. They show that every non-reduced model can be transformed into a reduced model by a sequence of linear embeddings.

**Proposition 2.11.** Let  $n \in \mathbb{N}_{>0}$ . Let  $\mathcal{M}$  be a model represented by  $(w_k, i_k, j_k)_{k=0}^n$ . If  $(i_l, j_l) = \mathbf{0}$  for some index l, then there exist a model  $\mathcal{M}'$ ,  $\lambda \in [0, 1]$  and  $k = 0, \ldots, n$  such that

$$\mathcal{M} = \Psi_{\lambda,k}(\mathcal{M}'),$$

where 
$$\Psi_{\lambda,k}: \Delta_{n-1} \to \Delta_n$$
 is defined as  $p_i \mapsto \begin{cases} \lambda p_i & \text{if } k \neq i, \\ 1 - \lambda & \text{if } k = i. \end{cases}$ 

*Proof.* Let  $(i_l, j_l) = \mathbf{0}$  for some index l. If  $w_l = 1$ , then  $w_m = 0$  for all  $m \neq l$ ; this contradicts  $w_m > 0$  by Proposition 2.1. Set  $\lambda = 1 - w_l > 0$  and k = l. Define the model  $\mathcal{M}'$  represented by

$$\left(\frac{w_h}{1-w_l}, i_h, j_h\right)_{h=0, h\neq l}^n.$$

Then,  $\mathcal{M} = \Psi_{\lambda,k}(\mathcal{M}')$ .

**Proposition 2.12.** Let  $n \in \mathbb{N}_{>0}$ . Let  $\mathcal{M}$  be model represented by  $(w_k, i_k, j_k)_{k=0}^n$ . If  $(i_m, j_m) = (i_l, j_l)$  for  $m \neq l$ , then there exist a model  $\mathcal{M}'$ ,  $\lambda \in [0, 1]$  and  $k, h = 0, \ldots, n$  such that

$$\mathcal{M} = \Psi_{\lambda,k,h}(\mathcal{M}'),$$

where 
$$\Psi_{\lambda,k,h}: \Delta_{n-1} \to \Delta_n$$
 is defined as  $p_i \mapsto \begin{cases} p_i & \text{if } i \notin \{k,h\}, \\ \lambda p_k & \text{if } k = i, \\ (1-\lambda)p_k & \text{if } h = i. \end{cases}$ 

*Proof.* Define  $\lambda = \frac{w_m}{w_m + w_l}$ , k = m, and h = l. Define the model  $\mathcal{M}'$  represented by

$$(w_g + \delta_{gm} w_l, i_g, j_g)_{q=0, q \neq l}^n$$
.

Then, 
$$\mathcal{M} = \Psi_{\lambda,k}(\mathcal{M}')$$
.

By repeatedly applying the two propositions, we can transform any model into a reduced model.

Corollary 2.13. If  $\Delta_n$  contains a model of degree d, then there also exists a reduced model of degree d in  $\Delta_m$  for some  $m \leq n$ .

#### 2.3 Fundamental Models

As before, models refer to one-dimensional discrete statistical models with rational MLE. The main building blocks for the classification of models are *fundamental models*; we will see that reduced models come from fundamental models.

**Definition 2.14.** We call a model represented by  $(w_k, i_k, j_k)_{k=0}^n$  fundamental if it is reduced and the equation  $p_0 + \dots p_n \equiv 1$  for given  $(i_k, j_k)_{k=0}^n$  uniquely determines the weights  $(w_k)_{k=0}^n$ .

**Example 2.15.** The binomial model with two trials is fundamental. Given  $(i_0, j_0) = (0, 2)$ ,  $(i_1, j_1) = (1, 1)$ , and  $(i_2, j_2) = (2, 0)$ , the equation  $p_0 + p_1 + p_2 = w_0 \theta^2 + w_1 \theta (1 - \theta) + w_2 (1 - \theta)^2 \equiv 1$  uniquely determines the weights  $w_0 = 1$ ,  $w_1 = 2$ ,  $w_2 = 1$ . To see this observe that this equation is equivalent to  $w_0 \theta^2 + w_1 \theta - w_1 \theta^2 + w_2 - w_2 2\theta + w_2 \theta^2 = 1$  which is equivalent to solving  $w_2 - 1 + \theta(w_1 - 2w_2) + \theta^2(w_0 - w_1 + w_2) = 0$  for all  $\theta \in \mathbb{R}$ .

**Example 2.16.** Consider the probability simplex  $\Delta_0$ . It only contains the model 1 which is fundamental.

**Example 2.17.** Now, consider the probability simplex  $\Delta_1$ . It only contains the models  $\theta \mapsto (\theta, 1 - \theta)$  and  $\theta \mapsto (1 - \theta, \theta)$  which are equivalent. They are fundamental.

We will see that fundamental models like the ones above are building blocks for all reduced models by *composition*.

**Definition 2.18.** Let  $\mathcal{M}$  and  $\mathcal{M}'$  be reduced models which are represented by functions  $f, g: \mathbb{Z}^2 \to \mathbb{R}_{\geq 0}$ , see Remark 2.10. Let  $\mu \in (0,1)$ . The *composite*  $\mathcal{M} *_{\mu} \mathcal{M}'$  of  $\mathcal{M}$  and  $\mathcal{M}'$  is the reduced model represented by the function

$$(i,j) \mapsto \mu f(i,j) + (1-\mu)g(i,j).$$

We are about to show that every reduced model is the composite of finitely many fundamental models.

**Proposition 2.19.** Let  $\mathcal{M}$  be a reduced model. Then  $\mathcal{M}$  is the composite of finitely many fundamental models.

*Proof.* For  $\Delta_0$  and  $\Delta_1$  we know that they only contain fundamental models, see Examples 2.16 and 2.17.

Assume we are given  $\Delta_n$  with  $n \geq 2$ , and let  $\mathcal{M}$  be a model that is not fundamental. We aim to show that  $\mathcal{M}$  can be expressed as a composite of two models,  $\mathcal{M}'$  and  $\mathcal{M}''$ , whose supports are proper subsets of supp( $\mathcal{M}$ ). Assume this is indeed the case. Then, by applying the same argument to  $\mathcal{M}'$  and  $\mathcal{M}''$ , we can recursively decompose each nonfundamental model into models with smaller supports. Since supp( $\mathcal{M}$ ) is finite, this recursive decomposition must eventually terminate, yielding a decomposition of  $\mathcal{M}$  into fundamental models. Thus, we have shown that any reduced model is the composite of a finite number of fundamental models.

Let us prove that  $\mathcal{M}$  is the composite of two models whose supports are proper subsets of  $\operatorname{supp}(\mathcal{M})$ . Since  $\mathcal{M}$  is not fundamental, the equation  $p_0 + \cdots + p_n = 1$  has distinct solutions  $\mathbf{w}, \mathbf{w}' \in \mathbb{R}^{n+1}_{>0}$ . Define  $\mathbf{v} := \mathbf{w} - \mathbf{w}' \neq \mathbf{0}$ . Then,

$$\sum_{k=0}^{n} v_k \theta^{i_k} (1 - \theta)^{j_k} = 0 \quad \forall \theta \in (0, 1).$$

Observe that there exist strictly positive and negative coefficients  $v_k$ . Define

$$\lambda := \min \left\{ \frac{w_k}{|v_k|} : k = 0, \dots, n, \ v_k < 0 \right\},$$

$$u_k := w_k + \lambda v_k \quad \text{for } k = 0, \dots, n,$$

$$S_1 := \{ (i_k, j_k) : k = 0, \dots, n, \ u_k \neq 0 \}.$$

Note that  $\lambda > 0$  since all the coefficients  $w_k$  are strictly positive by definition. Also observe that  $u_k \geq 0$  if  $v_k \geq 0$ . Moreover, by definition  $\frac{w_k}{|v_k|} \geq \lambda$  for all  $k \geq 0$ . Hence, if  $v_k < 0$ , we also have  $\frac{u_k}{v_k} = \frac{w_k}{v_k} + \lambda \leq 0$ . Multiplying by  $v_k < 0$  we obtain  $u_k \geq 0$ . All in all, we have  $u_k \geq 0$  for all  $k = 0, \ldots, n$ . Moreover,  $u_k = 0$  if and only if  $v_k < 0$  and  $\lambda = \frac{w_k}{|v_k|}$ . This shows that  $S_1 \subseteq \text{supp}(\mathcal{M})$ . Since  $u_0 + \ldots u_n = 1$ , we have found a reduced model  $\mathcal{M}'$  represented by  $(u_k, i_k, j_k)_{(i_k, j_k) \in S_1}$ .

For the second model, we define

$$\mu := \min \left\{ \frac{w_k}{u_k} : k = 0, \dots, n, \ u_k \neq 0 \right\},$$

$$t_k := \frac{w_k - \mu u_k}{1 - \mu} \quad \text{for } k = 0, \dots, n,$$

$$S_2 := \left\{ (i_k, j_k) : k = 0, \dots, n, \ t_k \neq 0 \right\}.$$

Similarly,  $\mu > 0$ . We have  $\mu < 1$  because some  $v_k$  is positive implying  $u_k > w_k$ . By definition, we have  $t_k \ge 0$ , and  $t_k = 0$  if and only if  $u_k \ne 0$  and  $\mu = \frac{w_k}{u_k}$ . This shows that  $S_2 \subsetneq \text{supp}(\mathcal{M})$ 

and  $S_1 \cup S_2 = \text{supp}(\mathcal{M})$ . Since  $t_0 + \cdots + t_n = 1$ , we have found a reduced model  $\mathcal{M}''$  represented by  $(t_k, i_k, j_k)_{(i_k, j_k) \in S_2}$ .

Finally, we see that  $w_k = \mu u_k + (1 - \mu)t_k$ . This shows that  $\mathcal{M} = \mathcal{M}' *_{\mu} \mathcal{M}''$ .

By applying the previous proposition with Corollary 2.13, we obtain the following corollary.

Corollary 2.20. If  $\Delta_n$  contains a non-fundamental model of degree d, then there exists a fundamental model of degree d in  $\Delta_m$  for some m < n.

**Example 2.21.** For the two-dimensional probability simplex  $\Delta_2$ , we can classify all models. Again, models refer to one-dimensional discrete statistical models with rational MLE. Note that the model  $\mathcal{M}$  parametrized by  $\theta \mapsto (\theta, 1 - \theta)$  satisfies  $\mathcal{M} *_{\mu} \mathcal{M} = \mathcal{M}$  for all  $\mu$ . Since  $\Delta_1$  only contains the model  $\theta \mapsto (\theta, 1 - \theta)$ , we can conclude that  $\Delta_2$  only contains fundamental models or models that are not reduced.

To find all the fundamental models in  $\Delta_2$ , we need to check for all sets  $S = \{(i_k, j_k)\}_{k=0}^2 \subset \mathbb{Z}_{>0}^2$  of size three if the equation  $p_0 + p_1 + p_2 = \sum_{k=0}^2 w_k \theta^{i_k} (1-\theta)^{j_k} = 1$  has a unique solution  $(w_0, w_1, w_2)$ . As we can see, a priori infinitely many sets S need to be checked. However, as we will see in the next section, only those sets S with max  $\{i+j: (i,j) \in S\} \leq 2n-1=3$  need to be considered. Clearly, this reduces the number of sets S to be checked to a finite number.

We compute that only the following supports uniquely determine the weights  $(w_0, w_1, w_2)$ :

$$\{(0,3),(1,1),(3,0)\},\{(0,2),(1,1),(2,0)\},\{(0,1),(1,1),(2,0)\},\{(0,2),(1,0),(1,1)\}.$$

They correspond to the fundamental models  $((1-\theta)^3, 3\theta(1-\theta), \theta^3)$ ,  $((1-\theta)^2, 2\theta(1-\theta), \theta^2)$ ,  $(1-\theta, \theta(1-\theta), \theta^2)$ , and  $((1-\theta)^2, \theta, \theta(1-\theta))$ . The fourth model is equivalent to the third model by a parametrization  $\theta \mapsto 1-\theta$  and permutation of the coordinates.

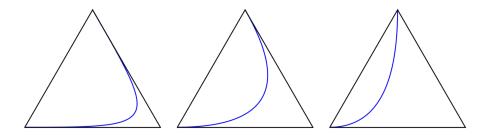


Figure 2.1: From left to right the illustration depicts the models parametrized  $((1-\theta)^3, 3\theta(1-\theta), \theta^3)$ ,  $((1-\theta)^2, 2\theta(1-\theta), \theta^2)$ ,  $(1-\theta, \theta(1-\theta), \theta^2)$ , and  $((1-\theta)^2, \theta, \theta(1-\theta))$ . The illustration is taken from [2].

We just computed all fundamental models of degree three or less in  $\Delta_2$ . We will see shortly that these are all models in the probability simplex  $\Delta_2$ . Of course,  $\Delta_2$  contains non-reduced models, too. These are models that come from linear embeddings  $\Psi_{\lambda,k}$  and  $\Psi_{\lambda,k,h}$ , see Proposition 2.11 and Proposition 2.12. There are infinitely many of them, and for  $\lambda = \frac{1}{3}$  we obtain the models  $\theta \mapsto (\frac{2}{3}\theta, \frac{1}{3}, \frac{2}{3}(1-\theta))$  and  $\theta \mapsto (1-\theta, \frac{1}{3}\theta, \frac{2}{3}\theta)$ .



Figure 2.2: This illustration depicts two non-reduced models in  $\Delta_2$  for  $\lambda = \frac{1}{3}$ . They are parametrized by  $\theta \mapsto (\frac{2}{3}\theta, \frac{1}{3}, \frac{2}{3}(1-\theta))$  and  $\theta \mapsto (1-\theta, \frac{1}{3}\theta, \frac{2}{3}\theta)$ . All other non-reduced models can be obtained by varying  $\lambda$ . The illustration is taken from [2].

Let us summarize the results of this section. It is the first part of our classification theorem.

**Theorem 2.22.** Every one-dimensional discrete statistical model with rational MLE in  $\Delta_n$  is the image of a reduced model in  $\Delta_m$  under a linear embedding  $\Delta_m \to \Delta_n$  for some  $m \le n$ .

Moreover, every reduced model  $\mathcal{M} \subset \Delta$  can be written as a composite of finitely many fundamental models

$$\mathcal{M} = \mathcal{M}_1 *_{\mu_1} (\cdots *_{\mu_{m-2}} (\mathcal{M}_{m-1} *_{\mu_{m-1}} \mathcal{M}_m))$$

for some m < n and  $\mu_1, \ldots, \mu_m \in (0, 1)$ .

*Proof.* See Proposition 2.19, Proposition 2.11, and Proposition 2.12.

### 2.4 On the Finiteness of Fundamental Models

We have established the first part of our classification theorem, namely that fundamental models are building blocks for all models. The second part is showing that there are only finitely many fundamental models in  $\Delta_n$  given  $n \in \mathbb{N}$ . Artuhr Bik and Orlando Marigliano proved that there are only finitely many fundamental models in  $\Delta_n$  for  $n \leq 4$  [2]. We will

later make significant progress towards proving the case n = 5. For  $n \ge 6$  no attempt has been made yet to the best of our knowledge.

Arthur Bik and Orlando Marigliano first proved the following proposition.

**Theorem 2.23.** Let  $\mathcal{M}$  be a one dimensional discrete statistical model with rational MLE in  $\Delta_n$ . For  $n \leq 4$  we have  $\deg(\mathcal{M}) \leq 2n - 1$ .

Given Theorem 2.23 it is easy to show the second part of our classification.

**Theorem 2.24.** There are only finitely many fundamental models in  $\Delta_n$  for all  $n \leq 4$ .

*Proof.* By Theorem 2.23 we know that the degree of a fundamental model is at most 2n-1. Since the number of supports of a fundamental model of degree 2n-1 is finite, there are only finitely many fundamental models in  $\Delta_n$  for all  $n \leq 4$ .

We will now spend the rest of this thesis on proving Theorem 2.23. The idea is to use the building blocks of fundamental models that we have established so far. Namely, it suffices to show the theorem for fundamental models.

**Theorem 2.25.** Let  $N \in \mathbb{N}$ . If for all  $n \leq N$  and for all fundamental models  $\mathcal{M} \in \Delta_n$  the upper bound  $\deg(\mathcal{M}) \leq 2n - 1$  holds, then the upper bound also holds for all statistical models in  $\Delta_{n'}$  for all  $n' \leq N$ .

Proof. Let  $N \in \mathbb{N}$  and  $n \leq N$ . Assume there is some non-fundamental model  $\mathcal{M}'$  in  $\Delta_n$  of degree greater than 2n-1. By Corollary 2.20 there exists a fundamental model  $\mathcal{M}$  in  $\Delta_m$  for some m < n of degree greater than 2m-1. This contradicts the assumption that the degree of fundamental models is at most 2n'-1 for all  $n' \leq N$ .

Counting all fundamental models in  $\Delta_n$  for  $n \leq 4$  is our guiding objective. As a first step, we introduce a combinatorial game that aids in counting fundamental models. We know that every reduced model can be represented by the sequence of triples  $(w_k, i_k, j_k)_{k=0}^n$ , where  $w_k \in \mathbb{R}_{>0}$  and  $i_k, j_k \in \mathbb{Z}_{\geq 0}$ . The model can be visualized in a directed graph with vertices in  $\mathbb{Z}^2$ , where we can place values  $w_k$  on vertices  $(i_k, j_k)$ . Each vertex (i, j) is connected by directed edges to (i + 1, j) and (i, j + 1).

Surprisingly, we can derive a combinatorial game from this graph by defining a specific set of rules. This game, called the *chipsplitting game*, will be rigorously introduced in the next chapter. After that, we will explore the game's properties and show how it can be used to count fundamental models in  $\Delta_n$  for  $n \leq 4$ .



Figure 2.3: The binomial model with two trials visualized in a directed graph with vertices in  $\{0,1,2,3\}^2$ .

## Chapter 3

## Chipsplitting Games

The notion of a chipsplitting game was introduced by [2] as a combinatorial approach to classifying one-dimensional discrete statistical models with rational maximum likelihood estimator. It was inspired by *chipfiring games* and for a subset of chipfiring games, the chipsplitting game is equivalent to the chipfiring game. We refer to [6] for a comprehensive introduction to chipfiring games.

#### 3.1 Basic Definitions

Let us define the notion of a chipsplitting game.

**Definition 3.1.** Let (V, E) be a directed graph without loops.

- (1) A chip configuration is a vector  $\mathbf{w} = (w_v)_{v \in V} \in \mathbb{Z}^V$  such that there are only finitely many nonzero components  $w_k$ .
- (2) The initial configuration is the chip configuration  $\mathbf{0} \in \mathbb{Z}^V$ .
- (3) A splitting move at  $u \in V$  maps a chip configuration  $\mathbf{w}$  to some chip configuration  $\mathbf{w}'$  defined by

$$w'_{v} := \begin{cases} w_{v} - 1 & \text{if } v = u, \\ w_{v} + 1 & \text{if } (u, v) \in E \\ w_{v} & \text{otherwise.} \end{cases}$$

This map is denoted by split,...

- (4) An unsplitting move at  $u \in V$  maps  $\mathbf{w}'$  back to  $\mathbf{w}$ . This map is denoted by unsplit<sub>u</sub>.
- (5) A chipsplitting game is a finite sequence of splitting and unsplitting moves.

- (6) An outcome of a chipsplitting game is the chip configuration obtained from applying the sequence of splitting and unsplitting moves defined by the game at the initial configuration.
- (7) Any outcome of a chipsplitting game is called an *outcome*.

**Proposition 3.2.** The order of the moves in a chipsplitting game does not affect the outcome.

*Proof.* This follows from commutativity of addition.

Note that all moves are reversible. Thus, we obtain the following corollary with Proposition 3.2.

Corollary 3.3. Let w be an outcome. Then, there exists a chipsplitting game whose outcome is w and where at no point both a splitting and an unsplitting move are applied at the same vertex.

Games that satisfy the condition in the corollary are called *reduced*. We will only consider reduced games in this thesis for simplicity. The map

{reduced games on 
$$(V, E)$$
} /  $\sim \rightarrow \{g : V' \to \mathbb{Z} : \#\{p \in V' : g(p) \neq 0\} < \infty\}$   
 $f \mapsto (p \mapsto \text{number of moves at } p \text{ in game } f)$ 

is a bijection, where  $V' \subset V$  is the subset of vertices with at least one outgoing edge. The equivalence relation  $\sim$  is defined by  $f \sim g$  if f and g are the same up to reordering. Unsplitting moves are counted negatively by  $p \mapsto$  number of moves at p in game f. Using the map above we identify a chipsplitting game with its corresponding function  $V' \to \mathbb{Z}$ . For every outcome  $\mathbf{w} = (w_v)_{v \in V}$  we have

$$w_v = -f(v) + \sum_{u \in V', (u,v) \in E} f(u),$$

where we define f(v) = 0 for  $v \notin V$ .

Now, we define the directed graphs that we will consider in this thesis. For  $d \in \mathbb{N} \cup \{\infty\}$  we write

$$V_d := \left\{ (i, j) \in \mathbb{Z}_{\geq 0}^2 \mid i + j \leq d \right\},$$
  
$$E_d := \left\{ (v, v + e) \mid v \in V_{d-1}, e \in \left\{ (1, 0), (0, 1) \right\} \right\}.$$

**Definition 3.4.** The degree  $deg(\mathbf{v})$  of a vertex  $\mathbf{v} = (i, j)$  is defined as i + j.

**Example 3.5.** A chip configuration  $\mathbf{w} = (w_{i,j})_{(i,j)\in V_d} \in \mathbb{Z}^{V_d}$  can be illustrated as a triangle of numbers where  $w_{i,j}$  is placed at the position (i,j) in the triangle. For example,  $w_{2,4} = 4$  means that the value 4 is placed in the second column and fourth row of the triangle. The following is an example of a sequence of chip configurations for d = 3:

When  $w_{i,j} = 0$ , we omit the value in the triangle and write a dot instead. The sequence above starts with the initial configuration and then applies a splitting move at the vertex (0,0), (1,0), (0,1), (0,2) and (2,0). Finally, we apply an unsplitting move at the vertex (1,1) to obtain the final configuration. Coming back to figure 2.4, we see that it is represented as the third configuration of the triangle above.

Let us define some more terminology.

**Definition 3.6.** Let  $\mathbf{w} = (w_{i,j})_{(i,j) \in V_d}$  be a chip configuration.

- (1) The positive support of  $\mathbf{w}$  is defined as  $\operatorname{supp}^+(\mathbf{w}) := \{(i, j) \in V_d \mid w_{i,j} > 0\}.$
- (2) The negative support of **w** is defined as supp<sup>-</sup>(**w**) :=  $\{(i,j) \in V_d \mid w_{i,j} < 0\}$ .
- (3) The support of  $\mathbf{w}$  is defined as the union of the positive and negative support.
- (4) The degree of **w** is defined as  $deg(\mathbf{w}) := max\{i + j \mid (i, j) \in supp(\mathbf{w})\}.$
- (5) We say **w** is valid if its negative support is empty or only contains (0,0).

We are interested in *outcomes* that are *valid* since they will correspond to reduced models as we will see later. For that reason, it would be convenient to have a criterion for when a chip configuration is an outcome. The next section will provide such a criterion with the help of *Pascal equations*.

**Example 3.7.** Consider the following chip configuration:

We clearly see that this configuration is valid, but is it also an outcome of a chipsplitting game? Currently, the only way to answer this question is to apply all possible sequences of splitting and unsplitting moves to the initial configuration and check if the outcome is the given configuration. In the next section, we present an easily computable characterization to answer this question.

## 3.2 Pascal Equations

In this chapter we will establish that outcomes are roots of Pascal equations. So let us first define Pascal equations which are special cases of *linear forms*.

**Definition 3.8.** A linear form on  $\mathbb{Z}^{V_d}$  is a map of the form

$$\mathbb{Z}^{V_d} \to \mathbb{Z}, \quad \mathbf{w} \mapsto \sum_{(i,j) \in V_d} c_{i,j} w_{i,j}.$$

It is denoted by  $\sum_{(i,j)\in V_d} c_{i,j} x_{i,j}$ .

**Definition 3.9.** A Pascal form on  $\mathbb{Z}^{V_d}$  is a linear form  $\sum_{(i,j)\in V_d} c_{i,j}x_{i,j}$  on  $\mathbb{Z}^{V_d}$  satisfying

$$c_{i,j} = c_{i+1,j} + c_{i,j+1}$$
 for all  $(i,j) \in V_{d-1}$ .

**Example 3.10.** We can visualize a Pascal form as a triangle of numbers where  $c_{i,j}$  is placed at the position (i,j) in the triangle. Here are examples of Pascal forms for d=2:

0			1			0			0	
1	1		1	0		0	0		1	1
2	1	0	1	0	0	1	1	1	0 .	-1 -2

Evaluating Pascal equations is invariant under splitting and unsplitting moves.

**Proposition 3.11.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a chip configuration. Let  $p = \sum c_{i,j} x_{i,j}$  be a Pascal equation on  $\mathbb{Z}^{V_d}$ . Then, we have  $p(\mathbf{w}) = p(\mathrm{split}_u(\mathbf{w})) = p(\mathrm{unsplit}_v(\mathbf{w}))$  for all  $u, v \in V_{d-1}$ .

*Proof.* Let  $u := (i', j') \in V_{d-1}$ . By the Pascal property, we have

$$c_{i'+1,j'} + c_{i',j'+1} - c_{i',j'} = 0.$$

Thus, we have

$$p(\text{split}_{u}(\mathbf{w})) = \sum_{(i,j)\in V_{d}} c_{i,j}(\text{split}_{u}(\mathbf{w}))_{i,j}$$

$$= \sum_{(i,j)\in V_{d}} c_{i,j} \begin{cases} w_{i,j} - 1 & \text{if } (i,j) = u, \\ w_{i,j} + 1 & \text{if } (i,j) \in \{(i'+1,j'), (i',j'+1)\} \\ w_{i,j} & \text{otherwise} \end{cases}$$

$$= \sum_{(i,j)\in V_{d}} c_{i,j}w_{i,j} = p(\mathbf{w}).$$

Similarly, we can show that  $p(\text{unsplit}_v(\mathbf{w})) = p(\mathbf{w})$  for all  $v \in V_{d-1}$ .

Corollary 3.12. Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be an outcome. Let  $p = \sum c_{i,j} x_{i,j}$  be a Pascal equation on  $\mathbb{Z}^{V_d}$ . Then,  $p(\mathbf{w}) = 0$ .

*Proof.* Clearly, we have  $p(\mathbf{0}) = 0$ . Then, we use Proposition 3.11 and the fact that  $\mathbf{w}$  is obtained from the initial configuration  $\mathbf{0}$  by a sequence of splitting and unsplitting moves.  $\square$ 

This demonstrates that outcomes are roots of Pascal equations. The converse is also true as we will see now. This is one of the most important results; so let us state it now.

**Theorem 3.13.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a chip configuration. Then,  $\mathbf{w}$  is an outcome if and only if  $\mathbf{w}$  is a root of all Pascal equations on  $\mathbb{Z}^{V_d}$ .

The direction left to right is the content of the previous corollary. For the other direction life would be easier if we had not to deal with infinitely many Pascal equations. So let us fix this first by introducing a basis from which we can generate all Pascal equations through linear combinations.

**Example 3.14.** Fix the degree d = 2. We later claim that the following set of Pascal forms is a basis:

Note that the first column of each Pascal form is a unit vector in  $\mathbb{R}^3$ . We can also fix the first row of each Pascal form to be a unit vector in  $\mathbb{R}^3$ :

We will denote the first set of Pascal forms by  $\{col(0), col(1), col(2)\}$  and the second set by  $\{row(0), row(1), row(2)\}$ .

To generalize the example above to an arbitrary degree  $d \in \mathbb{N}$  and to vectors beyond unit vectors, we assert that there exists a unique Pascal form whose first column is any chosen vector.

**Proposition 3.15.** Let  $\mathbf{a} = (a_0, \dots, a_d)$  be any vector with integer entries. Then, the following two statements hold:

- (1) There exists a unique Pascal form  $\sum c_{i,j}x_{i,j}$  such that  $c_{0,\cdot}=\mathbf{a}$ .
- (2) There exists a unique Pascal form  $\sum c_{i,j}x_{i,j}$  such that  $c_{\cdot,0}=\mathbf{a}$ .

*Proof.* Set  $c_{0,\cdot} := \mathbf{a}$ . Define  $c_{i+1,j} := c_{i,j} - c_{i,j+1}$  for all  $(i,j) \in V_d$  with i = 0. Then, we use the same formula to define  $c_{i+1,j}$  for all  $(i,j) \in V_d$  with i = 1. We repeat this process until we have defined all  $c_{i,j}$  for  $(i,j) \in V_d$ .

For the second statement, we set  $c_{\cdot,0} := \mathbf{a}$ . Define  $c_{i,j+1} := c_{i,j} - c_{i+1,j}$  for all  $(i,j) \in V_d$  with j = 0. Then, we use the same formula to define  $c_{i,j+1}$  for all  $(i,j) \in V_d$  with j = 1. We repeat this process until we have defined all  $c_{i,j}$  for  $(i,j) \in V_d$ .

Let us define our first two Pascal form bases.

**Definition 3.16.** Let k = 0, ..., d and  $\mathbf{e}_k \in \mathbb{R}^{d+1}$  be the k-th unit vector.

- We define col(k) to be the unique Pascal form  $\sum c_{i,j}x_{i,j}$  such that  $c_{0,\cdot}=\mathbf{e}_k$ .
- We define row(k) to be the unique Pascal form  $\sum c_{i,j}x_{i,j}$  such that  $c_{i,0}=\mathbf{e}_k$ .

For examples of the Pascal forms col(k) and row(k) for d=2 see Example 3.14. We provide another example for d=7.

**Example 3.17.** Let us consider the Pascal form col(3) for d = 7. We visualize this Pascal form as follows:

The Pascal form row(3) is visualized as follows:

```
-35
-20
     15
     10
        -4
-10
-4
     6
        -4
     3
-1
       -3
             1
     1
       -2
             1
        -1
             1
```

**Proposition 3.18.** For all integers k = 0, ..., d the following formulas hold:

$$col(k) = (-1)^k \sum_{(i,j)\in V_d} (-1)^j \binom{i}{k-j} x_{i,j},$$
$$row(k) = (-1)^k \sum_{(i,j)\in V_d} (-1)^i \binom{j}{k-i} x_{i,j}.$$

Note that  $\binom{a}{b} = 0$  for b < 0 or b > a.

*Proof.* We claim that  $(-1)^k \sum_{(i,j) \in V_d} (-1)^j {i \choose k-j} x_{i,j}$  is a Pascal equation. To see that observe

$$(-1)^{j} \binom{i+1}{k-j} + (-1)^{j+1} \binom{i}{k-j-1} = (-1)^{j} \binom{i}{k-j}$$

for all  $(i,j) \in V_d$  due to  $\binom{a}{b+1} + \binom{a}{b} = \binom{a+1}{b+1}$  where we set a=i and b=k-j-1. Next, we see that  $(-1)^{k+j} \binom{0}{k-j} = \delta_{jk}$ . Thus,  $(-1)^k \sum_{(i,j) \in V_d} (-1)^j \binom{i}{k-j} x_{i,j}$  is indeed  $\operatorname{col}(k)$ . By symmetry of the binomial coefficients, we can use the same argument to show the

second formula.

We now show that  $\{\operatorname{col}(k)\}_{k=0}^d$  is indeed a basis for all Pascal forms on  $\mathbb{Z}^{V_d}$ .

**Proposition 3.19.** Let p be a Pascal form on  $\mathbb{Z}^{V_d}$ . The following statements hold:

- (1) There exist unique coefficients  $\mu_0, \ldots, \mu_d \in \mathbb{Z}$  such that  $p = \mu_0 \operatorname{col}(0) + \cdots + \mu_d \operatorname{col}(d)$ .
- (2) There exist unique coefficients  $\lambda_0, \ldots, \lambda_d \in \mathbb{Z}$  such that  $p = \mu_0 \text{row}(0) + \cdots + \mu_d \text{row}(d)$ .

*Proof.* Let  $p = \sum c_{i,j} x_{i,j}$  be a Pascal form on  $\mathbb{Z}^{V_d}$ . If we try to solve the equation

$$\sum_{(i,j)\in V_d} c_{i,j} x_{i,j} = \lambda_0 \operatorname{col}(0) + \dots + \lambda_d \operatorname{col}(d)$$
(3.1)

for  $\lambda_0, \ldots, \lambda_d$ , then due to Proposition 3.18 we see for all  $(i, j) \in V_d$  that we have

$$c_{i,j} = \lambda_0 (-1)^{0+j} \binom{i}{0-j} + \lambda_1 (-1)^{1+j} \binom{i}{1-j} + \dots + \lambda_d (-1)^{d+j} \binom{i}{d-j}$$
$$= \lambda_j (-1)^{2j} \binom{i}{0} + \lambda_{j+1} (-1)^{2j+1} \binom{i}{1} + \dots + \lambda_{i+j} (-1)^{2j+i} \binom{i}{i}.$$

We see  $c_{0,\cdot} = (\lambda_0, \dots, \lambda_d)$ . Thus we set the coefficients  $\boldsymbol{\mu} \coloneqq c_{0,\cdot}$  and by Proposition 3.15 we see that  $\sum_{(i,j)\in V_d} c_{i,j}x_{i,j} = \mu_0\operatorname{col}(0) + \cdots + \mu_d\operatorname{col}(d)$ . Moreover, the same proposition shows that the coefficients  $\lambda_0, \ldots, \lambda_d$  in Equation 3.1 are uniquely determined.

For the second statement we use the same argument.

Corollary 3.20. The set  $\{\operatorname{col}(k)\}_{k=0}^d$  is a basis for all Pascal forms on  $\mathbb{Z}^{V_d}$ . The same holds for  $\{\operatorname{row}(k)\}_{k=0}^d$ .

*Proof.* This follows from the previous proposition.

Let us come back to Theorem 3.13. We can now prove the other direction; namely that roots of all Pascal equations on  $\mathbb{Z}^{V_d}$  are outcomes.

**Proposition 3.21.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a chip configuration. If for all Pascal equations p on  $\mathbb{Z}^{V_d}$  we have  $p(\mathbf{w}) = 0$ , then  $\mathbf{w}$  is an outcome.

*Proof.* Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a chip configuration. By assumption

$$\operatorname{col}(\operatorname{deg}(\mathbf{w}))(\mathbf{w}) = 0. \tag{3.2}$$

Note that by Proposition 3.18 for  $\operatorname{col}(\deg(\mathbf{w})) = \sum c_{i,j} x_{i,j}$  we have  $c_{i,\deg(\mathbf{w})-i} = (-1)^i$  for all  $i = 0, \ldots, \deg(\mathbf{w})$ . Moreover, we have

$$c_{i,j} = 0 \quad \text{for all } i + j < \deg(\mathbf{w})$$
 (3.3)

by Proposition 3.18. Together with Equation 3.2 and 3.3 we obtain

$$\sum_{i=0}^{\deg(\mathbf{w})} (-1)^i w_{i,\deg(\mathbf{w})-i} = 0.$$
(3.4)

Furthermore, we know that there exists a unique minimal set of splitting or unsplitting moves at vertices (i, j) of degree  $\deg(\mathbf{w}) - 1$  such that applied to  $\mathbf{w}$  we obtain a chip configuration  $\mathbf{w}'$  with  $w'_{i,j} = 0$  for all  $i = 0, ..., \deg(\mathbf{w})$ . We call applying these set of moves to  $\mathbf{w}$  retraction.

Thus,  $\mathbf{w}'$  has degree less than  $\deg(\mathbf{w})$ . By Proposition 3.11  $\mathbf{w}'$  is also a root of all Pascal equations. We repeat the retraction process  $\deg(\mathbf{w})$  many times until we obtain some chip configuration of degree 0. This chip configuration is the initial configuration due to Equation 3.4. Thus,  $\mathbf{w}$  is an outcome.

We have shown Theorem 3.13. Characterizing outcomes as roots of Pascal equations is a powerful tool to determine if a chip configuration is an outcome.

#### Algorithm 1 Validating outcomes

```
Require: chipsplitting configuration \mathbf{w} \in \mathbb{Z}^{V_d}
Ensure: True if w is an outcome, False otherwise
 1: function ISOUTCOME(A, n)
        initialize set S = \{ col(0), \ldots, col(deg(\mathbf{w})) \}
 2:
        for p of S do
 3:
 4:
            if p(\mathbf{w}) \neq 0 then
                return False
 5:
            end if
 6:
        end for
 7:
        return True
 9: end function
```

Proof of correctness of Algorithm 1. This follows from Theorem 3.13.

**Example 3.22.** Returning to Example 3.7, we see that the chip configuration is a root of all Pascal equations  $col(0), \ldots, col(6)$  using Algorithm 1. Thus, the chip configuration is an outcome.

#### 3.3 Valid Outcomes and Reduced Statistical Models

In the previous sections, we have established that outcomes are roots of Pascal forms. Now, we will demonstrate that a subset of *valid outcomes* are in one-to-one correspondence with

reduced statistical models. Thus, we obtain not only a combinatorial characterization of reduced statistical models through chip-splitting games but also an algebraic characterization through Pascal equations. As before, statistical models mean one-dimensional discrete statistical models with rational maximum likelihood estimator.

We remind that valid chipsplitting configurations are those where the negative support is empty or only contains the vertex (0,0). Hence, valid outcomes are roots of Pascal equations whose negative supports are empty or only contain the vertex (0,0).

The function  $\mathbf{w}(\mathcal{M})$  maps reduced models  $\mathcal{M} = (w_k, i_k, j_k)_{k=0}^n$  to chip configurations  $\mathbf{w}(\mathcal{M}) = (w_{i,j})_{(i,j) \in V_{\infty}}$  by

$$w_{i,j} := \begin{cases} -1 & \text{if } (i,j) = (0,0), \\ w_k & \text{if } (i,j) = (i_k, j_k) \text{ for some } k, \\ 0 & \text{otherwise.} \end{cases}$$

Note that the map  $\mathbf{w}(\mathcal{M})$  defines a *real* chipsplitting games; the rules of the game are the same as for integer chipsplitting games.

**Example 3.23.** The binomial model ((1,3,0),(3,2,1),(3,1,2),(1,0,3)) with three trials is mapped to the chip configuration below:

**Example 3.24.** Does the following valid real outcome from Example 3.7 induce a reduced statistical model through the inverse map  $\mathbf{w}^{-1}$ ?

The outcome would correspond to the reduced model

$$\mathcal{M} = ((0.5, 2, 0), (0.5, 4, 0), (2.5, 1, 3), (0.5, 1, 5), (4, 2, 1), (2.5, 2, 4),$$
$$(2.5, 3, 2), (1, 3, 3), (1, 5, 0), (1, 5, 1))$$

in the probability simplex  $\Delta_9$ . As it turns out  $\mathcal{M}$  is indeed a reduced statistical model by the next theorem.

**Theorem 3.25.** The map  $\mathcal{M} \mapsto w(\mathcal{M})$  is a bijection between reduced statistical models and valid real outcomes  $\mathbf{w} \in \mathbb{R}^{V_{\infty}}$  with  $w_{0,0} = -1$ .

reduced models « valid outcomes

Figure 3.1: Bijection between reduced models and valid real outcomes **w** with  $w_{0,0} = -1$ 

To show this theorem, we need to do some preparations. Let  $\mathbb{R}[\theta]_{\leq d}$  denote the vector space of polynomials in the variable  $\theta$  of degree at most d with real coefficients. Similarly, we define  $\mathbb{Z}[\theta]_{\leq d}$  and  $\mathbb{Q}[\theta]_{\leq d}$ . Next, we introduce the linear map  $\alpha_d^{\mathbb{R}}$  that maps real chip configurations to real polynomials:

$$\alpha_d^{\mathbb{R}} : \mathbb{R}^{V_d} \to \mathbb{R}[\theta]_{\leq d},$$
  

$$\mathbf{w} \mapsto \sum_{(i,j)\in V_d} w_{i,j} \theta^i (1-\theta)^j.$$

We define the map  $\alpha_d^{\mathbb{Z}}$  and  $\alpha_d^{\mathbb{Q}}$  for integer and rational chip configurations analogously.

**Lemma 3.26.** The following statements hold true for all  $d \in \mathbb{N} \cup \{\infty\}$ :

- (1)  $\{\mathbf{w} \in \mathbb{R}^{V_d} \mid \mathbf{w} \text{ is an outcome}\} = \text{kernel}(\alpha_d^{\mathbb{R}});$
- (2)  $\{\mathbf{w} \in \mathbb{Z}^{V_d} \mid \mathbf{w} \text{ is an outcome}\} = \text{kernel}(\alpha_d^{\mathbb{Z}});$
- (3)  $\{\mathbf{w} \in \mathbb{Q}^{V_d} \mid \mathbf{w} \text{ is an outcome}\} = \text{kernel}(\alpha_d^{\mathbb{Q}}).$

*Proof.* We only prove the first statement. The other two statements are proven analogously. Note that it suffices to show the statement for  $d < \infty$  since  $\alpha_{\infty}^{\mathbb{R}}$  is the direct limit of  $\alpha_0^{\mathbb{R}}, \alpha_1^{\mathbb{R}}, \alpha_2^{\mathbb{R}}, \ldots$ , and so on.

Let  $d < \infty$ . By Corollary 3.20, the codimension of the outcome space is d + 1, as it is defined by the roots of the Pascal forms  $col(0), \ldots, col(d)$ .

Let  $f(\theta) = \lambda_0 + \lambda_1 \theta + \cdots + \lambda_d \theta^d$  be a polynomial in  $\mathbb{R}$  of degree at most d. Define a chipsplitting configuration  $\mathbf{w}$  by

$$w_{i,j} \coloneqq \begin{cases} \lambda_i & \text{if } j = 0, \\ 0 & \text{otherwise.} \end{cases}$$

Then,  $\alpha_d^{\mathbb{R}}(\mathbf{w}) = f$ , which shows that the map  $\alpha_d^{\mathbb{R}}$  is surjective. Hence, the kernel of  $\alpha_d^{\mathbb{R}}$  has codimension d+1; it has equal codimension as the space of outcomes.

Finally, we just need to show that the space of outcomes is contained in the kernel of  $\alpha_d^{\mathbb{R}}$ . Since their codimensions are equal, the two spaces must be equal. Let  $\mathbf{w} \in \mathbb{R}^{V_d}$  be an

outcome. The value of  $\alpha_d^{\mathbb{R}}(\mathbf{w})$  remains the same if apply splitting or unsplitting moves at arbitrary vertices  $(i,j) \in V_{d-1}$  because we have

$$-\theta^{i}(1-\theta)^{j} + \theta^{i+1}(1-\theta)^{j} + \theta^{i}(1-\theta)^{j+1} = \theta^{i}(1-\theta)^{j}(-1+\theta+(1-\theta)) = 0.$$

The remaining claim follows from  $\alpha_d^{\mathbb{R}}(\mathbf{0}) = 0$ .

We now show Theorem 3.25.

Proof of Theorem 3.25. Let  $\mathcal{M} = (w_k, i_k, j_k)_{k=0}^n$  be a reduced model.

- First, we need to show that  $\mathbf{w} := w(\mathcal{M})$  is an outcome; a-priori we only know that it is some chip configuration. By definition of  $w(\mathcal{M})$ , we have that  $w_{0,0} = -1$ . Since  $\mathcal{M}$  is a statistical model, we know that  $\sum_{k=0}^{n} w_k \theta^{i_k} (1-\theta)^{j_k} \equiv 1$ . Thus,  $\alpha_d^{\mathbb{R}}(\mathbf{w}) = \sum_{k=0}^{n} w_k \theta^{i_k} (1-\theta)^{j_k} 1 \equiv 0$ . Thus,  $\mathbf{w} \in \text{kernel}(\alpha_d^{\mathbb{R}})$ . By Lemma 3.26, the chip configuration  $\mathbf{w}$  is an outcome.
- Injectivity: Let  $\mathcal{M} = (w_k, i_k, j_k)_{k=0}^n$  and  $\mathcal{M}' = (w'_k, i'_k, j'_k)_{k=0}^n$  be two distinct models. Then,  $w(\mathcal{M}) \neq w(\mathcal{M}')$  (see Remark 2.6).
- Surjectivity: Let  $\mathbf{w} \in \mathbb{R}^{V_{\infty}}$  be a valid real outcome with  $w_{0,0} = -1$ . We define

$$w_k := w_{i_k, j_k}$$
 for all  $k = 0, \dots, n$ .

Then,  $\mathcal{M} = (w_k, i_k, j_k)_{k=0}^n$  is a reduced model by Lemma 3.26. We see that  $w(\mathcal{M}) = \mathbf{w}$ . Hence,  $\mathcal{M} \mapsto w(\mathcal{M})$  is surjective.

Next, we collect simple results.

**Proposition 3.27.** The following statements hold for all reduced models  $\mathcal{M}$ :

- (1)  $\operatorname{supp}^+(w(\mathcal{M})) = \operatorname{supp}^+(\mathcal{M}).$
- (2) The map  $\mathcal{M} \mapsto w(\mathcal{M})$  is degree-preserving.
- (3) The outcome  $w(\mathcal{M})$  is a rational outcome if and only if all the coefficients of  $\mathcal{M}$  are rational.

*Proof.* All three statements follow directly from definitions.

**Proposition 3.28.** Let  $\mathbf{w} \in \mathbb{Q}^{V_{\infty}}$  be a valid rational outcome. Then, there exist positive  $\lambda \in \mathbb{Q}$  and integral valid outcome  $\mathbf{z} \in \mathbb{Z}^{V_{\infty}}$  such that  $\mathbf{w} = \lambda \mathbf{z}$ .

*Proof.* Let **w** be a valid rational outcome. Its support is finite. Thus, there exist  $\mu \in \mathbb{N}$  such that  $\mu \mathbf{w} \in \mathbb{Z}^{V_{\infty}}$ . Define  $\lambda := \frac{1}{\mu}$  and  $\mathbf{v} := \mu \mathbf{w}$ . Clearly,  $\alpha_{\deg(\mathbf{v})}^{\mathbb{Z}}(\mathbf{v}) = \mu \alpha_{\deg(\mathbf{w})}^{\mathbb{Q}}(\mathbf{w}) \equiv 0$ . By Lemma 3.26, **v** is an outcome. It is valid because  $\mu \mathbf{w}$  is valid.

**Proposition 3.29.** Let  $\mathbf{w} \in \mathbb{R}^{V_{\infty}}$  be a valid real outcome. If  $w_{0,0} = 0$ , then  $\mathbf{w} = \mathbf{0}$ .

*Proof.* By Lemma 3.26, we have  $\sum w_{i,j}\theta^i(1-\theta)^j\equiv 0$ . By assumption, the negative support is empty. Hence, all the  $w_{i,j}$  are non-negative. We evaluate at  $\theta=\frac{1}{2}$  to conclude that the positive support of  $\mathbf{w}$  is empty. Hence,  $\mathbf{w}=0$ .

Let us go back to Example 3.23.

**Example 3.30.** We have seen that the valid real outcome below induces a reduced statistical model by Theorem 3.25.

```
1 . 3 . . 3 . . 1.
```

The outcome has degree three and positive support size four. Can we find another outcome with the same degree but smaller positive support size? Indeed we can unsplit at vertex (1,1) to get

```
1
. .
. 3 .
-1 . . 1.
```

Can we reduce the positive support size further? As it will turn out, we cannot. The positive support size is minimal for a degree three outcome.

**Example 3.31.** Let us fix the positive support size to be three. What is the largest degree of a valid real outcome with positive support size three? We have already found

```
1
. .
. 3 .
-1 . . 1
```

However is there an even larger one? Once again, the answer is no: by the following theorem, the degree of a valid real outcome with positive support of size three is at most three.

**Theorem 3.32.** For valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| \leq 5$  we have  $\deg(\mathbf{w}) \leq 2 \cdot |\operatorname{supp}^+(\mathbf{w})| - 3$ .

Arthur Bik and Marigliano Orlando established this theorem in [2], and we will provide a proof in the following sections. The question of whether the upper bound extends to cases with larger positive support sizes,  $|\sup^+(\mathbf{w})| > 5$ , remains open. A primary contribution of this thesis is the significant advancement in proving this bound for the case  $|\sup^+(\mathbf{w})| = 6$ .

Pos. Support Size	Max Degree
2	1
3	3
4	5
5	7

Table 3.1: The following table shows the largest degree of a valid real outcome with positive support size 2, 3, 4 and 5.

Theorem 3.32 is of particular interest because it is equivalent to Theorem 2.23, i.e. proving that the degree of outcomes is bounded by their positive support size is equivalent to establishing that the degree of statistical models is bounded by their dimension.

degree of outcomes 
$$\mathbf{w} \longleftrightarrow$$
 degree of statistical models  $\mathcal{M}$   $\operatorname{supp}^+(\mathbf{w}) \longleftrightarrow$  dimension of  $\mathcal{M}$ 

**Proposition 3.33.** Theorem 2.23 and Theorem 3.32 are equivalent.

Proof. Assume Theorem 2.23 holds. We want to show Theorem 3.32. Let  $\mathbf{w}$  be a valid integral outcome of positive support size  $n \leq 5$ . Normalize  $\mathbf{w}$  such that  $w_{0,0} = -1$ . The degree and positive support size do not change. By Theorem 3.25, the outcome  $\mathbf{w}$  induces a reduced statistical model  $\mathcal{M}$  in  $\Delta_{n-1}$ . Then,  $\deg(\mathcal{M}) \leq 2(n-1) - 1 = 2n - 3$ . Thus, applying Theorem 3.25 let us go back to the outcome  $\mathbf{w}$ , and Proposition 3.27 establishes  $\deg(\mathbf{w}) \leq 2n - 3$ .

For the converse direction, assume Theorem 3.32 holds. Let  $n \leq 5$ . We want to show Theorem 2.23. Let  $\mathcal{M} = (w_k, i_k, j_k)_{k=0}^{n-1} \subset \Delta_{n-1}$  be a one-dimensional discrete statistical model with rational MLE. By Theorem 2.25 we may assume that  $\mathcal{M}$  is fundamental. We use Theorem 3.25 to map  $\mathcal{M}$  to some valid real outcome  $\mathbf{w} = (w_{i,j})$  with  $w_{0,0} = -1$ . Note that  $\mathbf{w}$  is even a rational outcome because by Definition 2.14 the weights  $(w_k)_{k=0}^{n-1}$  of the fundamental model  $\mathcal{M}$  are uniquely determined by the equation  $p_0(\theta) + p_1(\theta) + \cdots + p_{n-1}(\theta) - 1 \equiv 0$ , and for some  $\theta \in [0, 1]$  this equation becomes rational. Next, we use Proposition 3.28 to find some integral valid outcome  $\mathbf{z}$  such that  $\mathbf{w} = \mu \mathbf{z}$  for some positive scalar  $\mu \in \mathbb{Q}_{>0}$ . Again, scaling does not affect the degree or size of the positive support. By Proposition the integral valid outcome 3.27  $\mathbf{z}$  has positive support size n. By Theorem 3.32 we have  $\deg(\mathbf{z}) \leq 2n-3$ . Thus,  $\deg(\mathbf{w}) = \deg(\mu \mathbf{z}) = \deg(\mathbf{z}) \leq 2n-3$ . Hence, the degree of  $\mathcal{M}$  is smaller or equal to 2(n-1)-1 by Proposition 3.27. We proved Theorem 2.23.

Recall that our goal is to demonstrate that only finitely many fundamental statistical models exist in  $\Delta_n$  for  $n \leq 4$ . To achieve this, we originally aimed to prove Theorem 2.24 from Chapter 2. However, as established in Chapter 3, we can alternatively prove Theorem 3.32, which serves as an equivalent approach to resolving the problem. We have chosen to focus on Theorem 3.32 because it allows us to address the problem from a combinatorial perspective.

## Chapter 4

## Supports of Valid Outcomes

Let us devote the remaining chapters to the study of Theorem 3.32, which for the sake of convenience we restate below.

Theorem. The following upper bound

$$deg(\mathbf{w}) \le 2 \cdot |supp^+(\mathbf{w})| - 3$$

holds for valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| \le 5$ .

From now on, valid outcomes **w** refer to *integral* valid outcomes in  $\mathbb{Z}^{V_d}$  for *finite*  $d \in \mathbb{N}$ . To show the theorem, we will first study the supports of valid outcomes; knowing that some kinds of supports cannot be the supports of valid outcomes will help us to prove the theorem. For instance, integer configurations that have support in the entries below marked with an \* cannot be valid outcomes:

This is the key result of this chapter, see Section 4.5.

### 4.1 Invertibility Criterion

Let  $d \in \mathbb{N}$ . One of the most important tools in the study of outcomes is the *invertibility* criterion first introduced in [2]. By Theorem 3.13 we can characterize outcomes as the roots of all Pascal forms on  $\mathbb{Z}^{V_d}$ . In the previous chapter we have already found two bases for the

space of Pascal forms, namely (row(0), ..., row(d)) and (col(0), ..., col(d)) (see Definition 3.16). Let us introduce a *new* basis for the space of Pascal forms.

**Definition 4.1.** Let k = 0, ..., d and  $\mathbf{e}_k \in \mathbb{R}^{d+1}$  be the k-th unit vector. We define diag(k) to be the unique Pascal form  $\sum c_{i,j}x_{i,j}$  such that  $c_{k,d-k} = \mathbf{e}_k$ .

**Example 4.2.** Fix the degree d = 7. We visualized diag(3) by

**Proposition 4.3.** For all integers k = 0, ..., d we have:

$$\operatorname{diag}(k) = \sum_{(i,j)\in V_d} {\binom{d-i-j}{k-i}} x_{i,j}.$$

Note that  $\binom{a}{b} = 0$  for b < 0 or b > a.

*Proof.* Note that for all  $(i, j) \in V_d$  with i + j = d we have  $\binom{d-i-j}{k-i} = 1$  if and only if k = i, and in all other cases  $k \neq i$  the binomial coefficient is zero. Thus, it remains to show that  $\sum_{(i,j)\in V_d} \binom{d-i-j}{k-i} x_{i,j}$  is a Pascal form. We have

$$\binom{d-i-j}{k-i} = \binom{d-i-1-j}{k-i-1} + \binom{d-i-j-1}{k-i}.$$

for all  $(i,j) \in V_{d-1}$  because  $\binom{a+1}{b+1} = \binom{a}{b+1} + \binom{a}{b}$ .

**Proposition 4.4.** Let p be a Pascal form on  $\mathbb{Z}^{V_d}$ . There exist unique coefficients  $\mu_0, \ldots, \mu_d \in \mathbb{Z}$  such that  $p = \mu_0 \operatorname{diag}(0) + \cdots + \mu_d \operatorname{diag}(d)$ .

*Proof.* Let  $p = \sum c_{i,j} x_{i,j}$ . Choose  $\mu_k = c_{k,d-k}$  for  $k = 0, \dots, d$ . Since p is a Pascal form, the coefficients  $c_{i,j}$  satisfy the Pascal recurrence relation. Thus, the coefficients  $\mu_k$  are uniquely determined.

Given some set of vertices  $S \subset V_d$  the invertibility criterion uses the diagonal basis  $(\text{diag}(0), \ldots, \text{diag}(d))$  to determine whether a nonzero outcome with support in S exists.

**Definition 4.5.** Let  $E \subset \{0, \ldots, d\}$  and  $S \subset V_d$  with  $|E| = |S| \neq 0$ . The pairing matrix of (E, S) is definded as  $A_{E, S}^{(d)} := \left[\binom{d-i-j}{k-i}\right]_{k \in E, (i,j) \in S}$ .

**Example 4.6.** Let d = 2,  $S = \{(1, 1), (0, 0)\}$  and  $E = \{0, 1\}$ . Then the pairing matrix is

$$A_{E,S}^{(d)} = \begin{bmatrix} \binom{2-1-1}{0-1} & \binom{2-0-0}{0-2} \\ \binom{2-1-1}{1-1} & \binom{2-0-0}{1-2} \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix}.$$

Now, assume **w** is an outcome with support in S. Since it is an outcome, we have  $diag(k)(\mathbf{w}) = 0$  for all k = 0, 1, 2, 3. Thus,

$$A_{E,S}^{(d)}\mathbf{w} = \mathbf{0}.$$

We make the following observation: if the matrix  $A_{E,S}^{(d)}$  were invertible (it is not for the given example), then we would have  $\mathbf{w} = \mathbf{0}$ ; so in this case the initial configuration  $\mathbf{0}$  is the *only* outcome with support in S. This is the invertibility criterion.

**Proposition 4.7** (Invertibility Criterion). Let  $\mathbf{w}$  be an outcome with supp $(\mathbf{w}) \subset S$ . If  $A_{E,S}^{(d)}$  is invertible, then  $\mathbf{w} = \mathbf{0}$ .

Proof by Contraposition. Let  $\mathbf{w} \neq \mathbf{0}$ . Its support is non-empty. Then,  $\mathbf{w}' := (w_{i,j})_{(i,j) \in S} \neq \mathbf{0}$ . So,  $A_{E,S}^{(d)} \cdot \mathbf{w}' = \mathbf{0}$ . The kernel of the pairing matrix is non-trivial. Hence, the pairing matrix  $A_{E,S}^{(d)}$  is not invertible.

Given a non-zero configuration  $\mathbf{w}$  we try to construct sets  $S \supset \operatorname{supp}(\mathbf{w})$  and E such that the pairing matrix  $A_{E,S}^{(d)}$  is invertible. If we succeed, then  $\mathbf{w}$  is not an outcome since the initial configuration is the only valid outcome with support in S.

## 4.2 Divide and Conquer

The invertibility criterion is a powerful tool to determine whether a given configuration is an outcome. However, it is not always easy to find suitable sets S and E such that the pairing matrix is invertible. We will now introduce a method to construct such sets.

#### Divide

Instead of finding one large set S with  $\operatorname{supp}(\mathbf{w}) \subset S$ , we divide S into smaller sets  $S_1, \ldots, S_l$ . These smaller sets  $S_1, \ldots, S_k$  will be implicitly defined by integers  $\lambda_1, \ldots, \lambda_l \in \mathbb{N}$  as we will shortly see. We choose  $l \in \mathbb{N}$  and integers  $\lambda_1, \ldots, \lambda_l \in \mathbb{N}$  such that for all  $i = 1, \ldots, d$  we have

$$|S_i| \in \{0, \lambda_i\}$$

$$S_i := \{(i, j) \in \operatorname{supp}(\mathbf{w}) : i = c_{k-1}, \dots, c_k - 1\}$$

$$c_i := \lambda_1 + \dots + \lambda_i,$$

$$\lambda_1 + \dots + \lambda_l = d + 1.$$

**Remark 4.8.** The above decomposition works when  $|\{(i,j) \in \text{supp}(\mathbf{w}) : i \geq d-k\}| \leq k+1$  for all  $k = 0, \ldots, d$ . This is because we can always choose  $\lambda_1$  minimal such that  $|S_1| \in \{0, \lambda_1\}$ . We repeat this process until  $c_l = d+1$ . This decomposition is illustrated in the following example.

**Example 4.9.** Fix the degree d = 6. Assume we have some configuration  $\mathbf{w} \in \mathbb{Z}^{V_6}$  with support in the positions marked with an \* below.

The first column contains two non-zero entries. So we see  $\lambda_1 = 2$ . Then, we conclude that  $\lambda_2 = \lambda_3 = \lambda_4 = \lambda_5 = \lambda_6 = 1$ ; otherwise  $|S_i| \notin \{0, \lambda_i\}$  for i > 1.

Next with  $S_i$  defined, we define for all i = 1, ..., l the sets

$$E_i := \begin{cases} \{c_{i-1}, \dots, c_i - 1\} & \text{if } |S_i| = \lambda_i, \\ \emptyset & \text{if } |S_i| = 0. \end{cases}$$

We see that  $|E_i| = |S_i|$  for all i = 1, ..., l.

## Conquer

Given some support S, we divide it into smaller sets  $S_1, \ldots, S_l$  as described above. We also define sets  $E_1, \ldots, E_l$ . Write  $E := E_1 \cup \cdots \cup E_l$ .

Proposition 4.10. We have

$$A_{E,S}^{(d)} = \begin{bmatrix} A_{E_1,S_1}^{(d)} & 0 & \dots & 0 \\ A_{E_2,S_1}^{(d)} & A_{E_2,S_2}^{(d)} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ A_{E_l,S_1}^{(d)} & A_{E_l,S_2}^{(d)} & \dots & A_{E_l,S_l}^{(d)} \end{bmatrix}.$$

*Proof.* We have

$$A_{E,S}^{(d)} = \begin{bmatrix} A_{E_1,S_1}^{(d)} & A_{E_1,S_2}^{(d)} & \dots & A_{E_1,S_l}^{(d)} \\ A_{E_2,S_1}^{(d)} & A_{E_2,S_2}^{(d)} & \dots & A_{E_2,S_l}^{(d)} \\ \vdots & \vdots & \ddots & \vdots \\ A_{E_l,S_1}^{(d)} & A_{E_l,S_2}^{(d)} & \dots & A_{E_l,S_l}^{(d)} \end{bmatrix}.$$

Let x, y = 1, ..., l such that x < y. Let  $k \in E_x$  and  $(i, j) \in S_y$ . Then,  $k \le c_x - 1 < c_x \le c_{y-1} \le i$ ; so k - i < 0. Thus,  $\binom{d-i-j}{k-i} = 0$ . This implies that the upper off-diagonal blocks are zero.

Corollary 4.11. The pairing matrix  $A_{E,S}^{(d)}$  is invertible if and only if  $A_{E_1,S_1}^{(d)}, \ldots, A_{E_l,S_l}^{(d)}$  are invertible.

Corollary 4.12 (Invertibility Criterion, Divide and Conquer). Let  $\mathbf{w}$  be an outcome with  $\operatorname{supp}(\mathbf{w}) \subset S$ . If  $A_{E_1,S_1}^{(d)}, \ldots, A_{E_l,S_l}^{(d)}$  are invertible, then  $\mathbf{w} = \mathbf{0}$ .

**Example 4.13.** We continue Example 4.9.

With  $\lambda = (2, 1, 1, 1, 1, 1)$  we obtain the following decomposition

$$S_1 = \{(0,0),(0,4)\}, S_2 = \{(2,0)\}, S_3 = \emptyset, S_4 = \{(4,1)\}, S_5 = \{(5,0)\}, S_6 = \{(6,0)\}, S_6 = \{(6,0)\}, S_6 = \{(6,0)\}, S_6 = \{(6,0)\}, S_7 = \{(6,0)\}, S_8 = \{(6,0)\}, S_8$$

and

$$E_1 = \{0, 1\}, E_2 = \{2\}, E_3 = \emptyset, E_4 = \{4\}, E_5 = \{5\}, E_6 = \{6\}.$$

Then, the pairing matrix reads

$$A_{E,S}^{(d)} = \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ 6 & 2 & 0 & 0 & 0 & 0 \\ * & * & 1 & 0 & 0 & 0 \\ * & * & * & 1 & 0 & 0 \\ * & * & * & * & 1 & 0 \\ * & * & * & * & * & 1 \end{bmatrix},$$

where \* stands for arbitrary entries. The matrix is invertible, so no nonzero outcome with support in  $S = \{(0,0), (0,4), (2,0), (4,1), (5,0), (6,0)\}$  exists.

## 4.3 Implementation of the Hyperfield Criterion

The Invertibility Criterion inspires the following algorithm for determining if a set of indices  $S \subset V_d$  can actually be the support of some nonzero outcome in  $\mathbb{Z}^{V_d}$ . It is later used in the proof of Theorem 7.5 and Proposition 9.2.

#### Algorithm 2 Only Zero Outcome

**Require:** Positive support set  $S \subset V_d$ 

**Ensure:** True only if **0** is the only outcome **w** with supp<sup>+</sup>(**w**)  $\subset S$ , False if unconclusive

- 1:  $support \leftarrow \{(0,0)\} \cup S$
- 2:  $E \leftarrow \{0, \dots, |\mathtt{support}| 1\}$
- 3:  $P \leftarrow build\_pairing\_matrix(d, E, support)$
- 4: return rank(P)  $\neq$  |support|

The function build\_pairing\_matrix constructs the pairing matrix  $A_{E,S}^{(d)}$  as defined in Definition 4.5. If this Algorithm is inconclusive, we can try to use another set  $E \subset \{0, \ldots, d\}$ . This leads to the following generalization:

#### Algorithm 3 Only Zero Outcome (Generalized)

**Require:** Positive support set  $S \subset V_d$ , set  $E \subset \{0, \ldots, d\}$  with size |S| + 1

**Ensure:** True only if **0** is the only outcome **w** with supp<sup>+</sup>(**w**)  $\subset S$ , False if unconclusive

- 1:  $support \leftarrow \{(0,0)\} \cup S$
- 2:  $P \leftarrow build\_pairing\_matrix(d, E, support)$
- 3: return rank(P)  $\neq$  |support|

Algorithm Only Zero Outcome is clearly a special case of Algorithm Only Zero Outcome (Generalized) with  $E = \{0, ..., |S|\}$ .

## 4.4 Symmetry

With the Invertibility Criterion we can exclude certain supports from being the supports of valid outcomes. We will now show that the supports of valid outcomes are symmetric with respect to the main diagonal.

**Proposition 4.14.** Let  $\mathbf{w} = (w_{i,j})_{(i,j) \in V_d}$  be a configuration in  $\mathbb{Z}^{V_d}$ . Then  $\mathbf{w}$  is an outcome if and only if  $\tilde{\mathbf{w}} := (w_{j,i})_{(i,j) \in V_d}$  is an outcome.

*Proof.* Observe that

$$\operatorname{diag}(k) = \sum_{(i,j) \in V_d} \binom{d-i-j}{k-i} x_{i,j} = \sum_{(i,j) \in V_d} \binom{d-i-j}{d-i-j-(k-i)} x_{i,j} = \sum_{(i,j) \in V_d} \binom{d-i-j}{d-k-j} x_{i,j}.$$

Let **w** be a valid outcome. Then,  $\operatorname{diag}(k)(\mathbf{w}) = 0$  for all  $k = 0, \dots, d$ . Thus,

$$\operatorname{diag}(k)(\tilde{\mathbf{w}}) = \sum_{(i,j)\in V_d} {d-i-j \choose k-j} w_{j,i}$$

$$= \sum_{(i,j)\in V_d} {d-i-j \choose d-k-i} w_{j,i}$$

$$= \sum_{(i,j)\in V_d} {d-i-j \choose d-k-j} w_{i,j}$$

$$= \sum_{(i,j)\in V_d} {d-i-j \choose k-i} w_{i,j}$$

$$= \operatorname{diag}(k)(\mathbf{w}) = 0 \quad \forall k = 0, \dots, d.$$

**Example 4.15.** We have already seen that the support below cannot be the support of a valid outcome.

By symmetry, the support below cannot be the support of a valid outcome either.

Next, we introduce another kind of symmetry. Let  $\mathbf{w} = (w_{i,j})$  be a configuration in  $\mathbb{Z}^{V_d}$ . We define

$$\mathbf{w} \mapsto \hat{\mathbf{w}} := ((-1)^{d-j} w_{j,d-i-j})_{(i,j) \in V_d}$$

**Proposition 4.16.** Let  $\mathbf{w} = (w_{i,j})_{(i,j) \in V_d}$  be a configuration in  $\mathbb{Z}^{V_d}$ . Then  $\mathbf{w}$  is an outcome if and only if  $\hat{\mathbf{w}} := ((-1)^{d-j}w_{j,d-i-j})_{(i,j) \in V_d}$  is an outcome.

*Proof.* Let  $k = 0, \ldots, d$ . We have

$$col(k)(\hat{\mathbf{w}}) = (-1)^k \sum_{(i,j)\in V_d} (-1)^j \binom{i}{k-j} (-1)^{d-j} w_{j,d-i-j}$$
$$= (-1)^{d-k} \sum_{(i,j)\in V_d} \binom{d-i-j}{k-i} w_{i,j}$$
$$= (-1)^{d-k} \operatorname{diag}(k)(\mathbf{w}) = 0.$$

The symmetry just introduced can be interpreted in the following way: we define a group action of the symmetry group  $S_3$  on  $\mathbb{Z}^{V_d}$  by

$$(12) \cdot \mathbf{w} = \tilde{\mathbf{w}} \quad \text{and} \quad (123) \cdot \mathbf{w} = \hat{\mathbf{w}}.$$

Then, the group actions (12), (13) and (23) can be depicted in Figure 4.1.

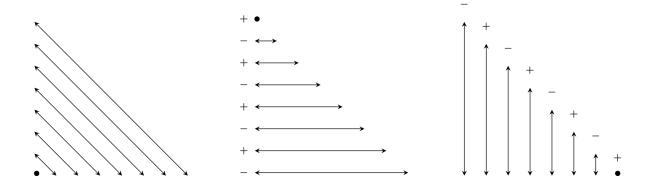


Figure 4.1: This illustration is taken from [2]. The left most illustration shows mirroring the configuration with respect to the main diagonal. The middle illustration shows switching the order on the same row while also alternating the signs of the row. The right most illustration shows switching the order on the same column while also alternating the signs of the column.

## 4.5 Impossible Supports

Now, we show that specific supports cannot be the supports of valid integral outcomes. Hence, the title of this section *Impossible Supports*. For instance, can we have an outcome

whose support is only contained in  $S = \{(0,0), (0,i)\}$  for some  $i \in \mathbb{N}$ ? We will show that this is not possible.

**Proposition 4.17.** Let  $d \in \mathbb{N}$ , and i = 0, ..., d. If  $S = \{(0, i)\}$  and  $E = \{0\}$ , then  $A_{E,S}^{(d)}$  is invertible.

*Proof.* We have  $A_{E,S}^{(d)} = [1]$ , which is invertible.

**Proposition 4.18.** Let  $d \in \mathbb{N}$ . Assume i, j = 0, ..., d with i < j. If  $S = \{(0, i), (0, j)\}$  and  $E = \{0, 1\}$ , then  $A_{E,S}^{(d)}$  is invertible.

*Proof.* We have 
$$A_{E,S}^{(d)} = \begin{bmatrix} 1 & 1 \\ d-i & d-j \end{bmatrix}$$
, which is invertible.

**Proposition 4.19.** Let  $d \in \mathbb{N}$ . Assume i, j, k = 0, ..., d with i < j < k. If  $S = \{(0, i), (0, j), (0, k)\}$  and  $E = \{0, 1, 2\}$ , then  $A_{E,S}^{(d)}$  is invertible.

*Proof.* We have

$$A_{E,S}^{(d)} = \begin{bmatrix} \binom{d-i}{0} & \binom{d-j}{0} & \binom{d-k}{0} \\ \binom{d-i}{0} & \binom{d-j}{0} & \binom{d-k}{0} \\ \binom{d-i}{1} & \binom{d-j}{2} & \binom{d-k}{0} \\ \binom{d-j}{2} & \binom{d-k}{2} & \binom{d-k}{2} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ d-i & d-j & d-k \\ \frac{(d-i)(d-i-1)}{2} & \frac{(d-j)(d-j-1)}{2} & \frac{(d-k)(d-k-1)}{2} \end{bmatrix}.$$

We substitute

$$x = d - i,$$
  

$$y = d - j,$$
  

$$z = d - k.$$

Then, we see

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 2 \end{bmatrix} A_{E,S}^{(d)} = \begin{bmatrix} 1 & 1 & 1 \\ x & y & z \\ x^2 & y^2 & z^2 \end{bmatrix}.$$

The matrix on the right-hand side is invertible because it is a Vandermonde matrix. Thus, the pairing matrix  $A_{E,S}^{(d)}$  is invertible.

**Proposition 4.20.** Let  $d \in \mathbb{N}$ . Assume i, j = 0, ..., d with i < j. Moreover, let k = 0, ..., d - 1. If  $S = \{(0, i), (0, j), (1, k)\}$ ,  $E = \{0, 1, 2\}$ , and  $i + j \neq 2k + 1$ , then  $A_{E,S}^{(d)}$  is invertible.

*Proof.* We have

$$A_{E,S}^{(d)} = \begin{bmatrix} \binom{d-i}{0} & \binom{d-j}{0} & \binom{d-k-1}{-1} \\ \binom{d-i}{0} & \binom{d-j}{0} & \binom{d-k-1}{0} \\ \binom{d-i}{2} & \binom{d-j}{2} & \binom{d-k-1}{1} \end{bmatrix} = \begin{bmatrix} 1 & 1 & 0 \\ d-i & d-j & d-k-1 \\ \frac{(d-i)(d-i-1)}{2} & \frac{(d-j)(d-j-1)}{2} & \frac{(d-k-1)(d-k-2)}{2} \end{bmatrix}.$$

We substitute

$$x = d - i,$$
  

$$y = d - j,$$
  

$$z = d - k - 1.$$

Then, we see

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 1 & 2 \end{bmatrix} A_{E,S}^{(d)} \begin{bmatrix} 1 & 1 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{1}{x-y} & 0 \\ 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ x & 1 & 1 \\ x^2 & x+y & 2z+1 \end{bmatrix}.$$

We see that the determinant is nonzero because  $x + y \neq 2z + 1$  by  $i + j \neq 2k + 1$ .

Remark 4.21. Without loss of generality, we may assume

$$S \subset \{(i, j) \in V_d \mid i < |S|\}$$
  
 $E = \{0, 1, \dots, |S| - 1\}$ 

because the matrices  $A_{E,S}^{(d)} = A_{E-\rho,S-\rho}^{(d-\rho)}$  are equal, where  $\rho := \min\{E \cup \{i \mid (i,j) \in S\}\}$  and  $E-\rho := \{(i-\rho,j) \mid (i,j) \in E\}$ . This assumption allows us to apply the previous propostions to more general S and E.

**Example 4.22.** Assume we have a configuration with support  $S = \{(0, i), (0, j), (0, k)\}$  for  $0 \le i < j < k \le d$ . By Proposition 4.19, we know that no valid integral nonzero outcome can have this support.

Now, let us consider a configuration  $\mathbf{w}$  with support supp $(\mathbf{w}) \subset S$  such that S can be decomposed into  $S_1, \ldots, S_l$  as described before. Let  $\ell = 1, \ldots, l$ . If  $S_\ell = \{(x, i), (x, j), (x, k)\}$  for  $0 \le i < j < k \le d$  and  $x \in \mathbb{N}$ , then  $\mathbf{w} = \mathbf{0}$  by Proposition 4.19 and the previous comment on the generality of S and E.

For instance, this configuration is not an outcome

where \* denotes a non-zero entry. This is because for  $\lambda = (3,3,1,1)$  we have  $S_2 = \{(3,0),(3,1),(3,3)\}$ . Similarly, these configurations are not outcomes

Of course, there are many more examples of impossible supports.

# Chapter 5

# Valid Outcomes of Positive Support Size One, Two, and Three

All the tools are ready to show the following three theorems, which were first proved in [2].

**Theorem 5.1.** No valid integral outcomes of positive support size one exists.

**Theorem 5.2.** For valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| = 2$  we have  $\deg(\mathbf{w}) = 1$ .

**Theorem 5.3.** For valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| = 3$  we have  $\deg(\mathbf{w}) \leq 3$ .

This proves our Main Theorem 3.32 for the case of positive support size three or less, i.e.

$$\deg(\mathbf{w}) \le 2 \cdot |\mathrm{supp}^+(\mathbf{w})| - 3$$

for all valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| \leq 3$ .

We start with the proof of the first theorem.

Proof of Theorem 5.1. Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid integral outcome. Since it is valid, we either have an empty negative support or a negative support that only contains (0,0). If the negative support is empty, then  $\mathbf{w} = \mathbf{0}$  by Proposition 3.29. Hence, we assume  $w_{0,0} < 0$ .

Now, consider the Pascal form diag(0) =  $\sum c_{i,j}x_{i,j}$ . We have  $c_{0,0} = c_{0,1} = \cdots = c_{0,d} = 1$  and  $c_{i,j} = 0$  for everything else. Similarly, we have for the Pascal form diag(d) =  $\sum c'_{i,j}x_{i,j}$  that  $c'_{\cdot,0} = \mathbf{1}$  and  $c'_{i,j} = 0$  for everything else. Since outcomes are roots of Pascal forms, we have

$$\operatorname{diag}(0)(\mathbf{w}) = \operatorname{diag}(d)(\mathbf{w}) = 0.$$

Since  $w_{0,0} < 0$  we must have  $w_{0,j} > 0$  and  $w_{i,0} > 0$  for some i, j > 0. Hence, **w** has positive support size at least two.

Next, we prove the second theorem.

# CHAPTER 5. VALID OUTCOMES OF POSITIVE SUPPORT SIZE ONE, TWO, AND THREE

Proof of Theorem 5.2. Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be an integral outcome with positive support size two and degree d. By the previous proof, we see that

$$supp^{+}(\mathbf{w}) = \{(0, j), (i, 0)\}.$$

Without loss of generality, we assume i = d. We want to show that j = d. Consider the Pascal form  $row(d) = \sum c_{i,j} x_{i,j}$ , which has only nonzero coefficients  $c_{i,j}$  for i + j = d.

- If the degree d is odd, we have  $c_{d,0} = 1$  and  $c_{0,d} = -1$ . Since  $row(d)(\mathbf{w}) = 0$ , we must have j = d.
- If the degree d is even, we have  $c_{d,0} = c_{0,d} = 1$ . Thus,  $row(d)(\mathbf{w}) \neq 0$  for all  $j = 0, \ldots, d$ . Hence, valid outcomes with positive support size two do not exist for even degrees.

From now on, we assume  $\operatorname{supp}^+(\mathbf{w}) = \{(0, d), (d, 0)\}$ . We could further assume that the degree d is odd, but we do not need it.

For sake of contradiction, let  $d \geq 2$ . Then, we can divide the support

$$supp(\mathbf{w}) = \{(0,0), (0,d), (d,0)\},\$$

via  $\lambda = (2, 1, ..., 1)$  as in Chapter 4.2 to obtain  $S_1 = \{(0, 0), (0, d)\}$ ,  $S_k = \emptyset$ , and  $S_l = \{(d, 0)\}$ . By Proposition 4.18, the pairing matrix induced by  $S_1$  and  $E_1 = \{0, 1\}$  is invertible. For  $S_l$  we apply Proposition 4.17 and Remark 4.21 to get that the induced pairing matrix is invertible. By Corollary 4.12, the outcome  $\mathbf{w}$  is zero, which has an empty positive support. This is a contradiction to the assumption that the positive support size is two. Hence, the degree d equals one.

**Example 5.4.** The previous theorem shows that the only valid integral outcomes with positive support size two are multiples of

It remains to prove Theorem 5.3. For that consider the following lemma which characterizes the possible supports of valid integral outcomes with positive support size three.

**Proposition 5.5.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid integral outcome of degree d. If the positive support size of  $\mathbf{w}$  is three, then one of the following holds:

- (1) We have  $supp(\mathbf{w}) = \{(0,0), (d,0), (0,d), (i,j)\}\$ for some i,j > 0 with i+j < d.
- (2) We have  $supp(\mathbf{w}) = \{(0,0), (d,0), (0,d), (i,d-i)\}\$  for some  $i = 1, \dots, d-1$ .
- (3) We have  $supp(\mathbf{w}) = \{(0,0), (d,0), (0,d), (i,0)\}\$ for some  $i = 1, \dots, d-1$ .

CHAPTER 5. VALID OUTCOMES OF POSITIVE SUPPORT SIZE ONE, TWO, AND THREE  $\phantom{a}$ 

- (4) We have  $supp(\mathbf{w}) = \{(0,0), (d,0), (0,d), (0,i)\}\$ for some  $i = 1, \dots, d-1$ .
- (5) We have  $supp(\mathbf{w}) = \{(0,0), (d,0), (0,e), (d-f,f)\}\$ for some  $e, f = 1, \dots, d-1$ .
- (6) We have  $supp(\mathbf{w}) = \{(0,0), (0,d), (e,0), (d-f,f)\}\$  for some  $e, f = 1, \dots, d-1$ .

*Proof.* Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid integral outcome of degree d. Assume  $\{(0,0),(d,0),(0,d)\} \subset \text{supp}(\mathbf{w})$ . Clearly, statement 1, 2, 3, or 4 must hold.

So assume  $(0,d) \notin \operatorname{supp}(\mathbf{w})$  and  $(d,0) \notin \operatorname{supp}(\mathbf{w})$ . As in the proof of Theorem 5.1, consider the Pascal form  $\operatorname{diag}(0) = \sum c_{i,j} x_{i,j}$ . We have  $c_{0,\cdot} = \mathbf{1}$  and  $c_{i,j} = 0$  for everything else. Similarly, we have for the Pascal form  $\operatorname{diag}(d) = \sum c'_{i,j} x_{i,j}$  that  $c'_{\cdot,0} = \mathbf{1}$  and  $c'_{i,j} = 0$  for everything else. Since outcomes are roots of Pascal forms, we have  $\operatorname{diag}(0)(\mathbf{w}) = 0$  diag $(d)(\mathbf{w}) = 0$ . Due to  $w_{0,0} < 0$ , we conclude  $w_{0,j} > 0$  and  $w_{i,0} > 0$  for some i, j > 0. Thus, we have

$$\{(i,0),(0,j)\}\subset \operatorname{supp}(\mathbf{w})$$

for some i, j = 1, ..., d-1 using the assumption  $(0, d) \notin \text{supp}(\mathbf{w})$  and  $(d, 0) \notin \text{supp}(\mathbf{w})$ . Since  $\mathbf{w}$  is of degree d, there exists  $w_{k,d-k} > 0$  for some k = 1, ..., d-1. However,  $\text{row}(d)(\mathbf{w}) = 0$  implies that there must be some  $w_{h,d-h} > 0$  for some  $h \neq k$ ; this h cannot equal 0 or d. Thus, the positive support size of  $\mathbf{w}$  is at least four, which is a contradiction. Hence, we must have  $(d, 0) \in \text{supp}(\mathbf{w})$  or  $(0, d) \in \text{supp}(\mathbf{w})$ .

Let  $(d,0) \in \text{supp}(\mathbf{w})$  and  $(e,0) \in \text{supp}(\mathbf{w})$  for some  $e=1,\ldots,d-1$ . Now using the same argument as before, there must exist some  $w_{f,d-f} > 0$  for some  $f=1,\ldots,d-1$ ; otherwise  $\text{row}(d)(\mathbf{w}) > 0$  which is a contradiction since  $\mathbf{w}$  is a root of all Pascal forms. This proves statement 5.

The proof for statement 6 is analogous.

Knowing the possible supports of valid integral outcomes with positive support size three, we apply the Invertibility Criterion 4.12 to each possible support to prove Theorem 5.3.

**Proposition 5.6.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid integral outcome of degree d. If  $\operatorname{supp}(\mathbf{w}) = \{(0,0),(d,0),(0,d),(i,j)\}$  for some i,j>0 with i+j< d, then d=3 and (i,j)=(1,1).

Proof. Let i > 1. Choose  $\lambda = (2, 1, ..., 1)$ . Then,  $E_1 = \{0, 1\}$ ,  $S_1 = \{(0, 0), (0, d)\}$ ,  $E_{i-1} = \{i\}$ ,  $S_{i-1} = \{(i, j)\}$ ,  $E_{d-1} = \{d\}$ ,  $S_{d-1} = \{(d, 0)\}$  and,  $E_k = S_k = \emptyset$  for all  $k \in \{1, ..., d-1\} \setminus \{1, i-1, d-1\}$ . The pairing matrices  $A_{E_n, S_n}^{(d)}$  are all invertible for n = 1, ..., d-1. Hence, the pairing matrix  $A_{\{0,1,i,d\}, \text{supp}(\mathbf{w})}^{(d)}$  is also invertible. By the Invertibility Criterion,  $\mathbf{w}$  is the zero configuration, which is a contradiction. Thus, we have i = 1.

Now, we assume j > 1. The configuration  $\tilde{\mathbf{w}} = (w_{ji})_{(i,j) \in V_d}$  is an outcome by Proposition 4.14 because  $\mathbf{w}$  is an outcome. Then  $\tilde{\mathbf{w}}$  has support  $\{(0,0),(d,0),(0,d),(1,\cdot)\}$  by the previous argument. However, then we have j = 1 which is a contradiction. So, we have j = 1.

Finally, we need to show that the degree d equals three. For the sake of contradiction, assume d > 3. Then, we can choose  $\lambda = (3, 1, ..., 1)$ . We obtain  $E_1 = \{0, 1, 2\}$  and

CHAPTER 5. VALID OUTCOMES OF POSITIVE SUPPORT SIZE ONE, TWO, AND THREE

 $S_1 = \{(0,0), (0,d), (1,1)\}$ . By Proposition 4.20 this pairing matrix  $A_{E_1,S_1}^{(d)}$  is invertible. The other relevant pairing matrix  $A_{\{d\},\{(d,0)\}}^{(d)}$  is also invertible. Thus, the pairing matrix  $A_{\{0,1,2,d\},\text{supp}(\mathbf{w})}^{(d)}$  is invertible. By the Invertibility Criterion, the configuration  $\mathbf{w}$  is the zero configuration, which is a contradiction. Hence, we have d = 3.

**Proposition 5.7.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid integral outcome of degree d. Assume the outcome  $\mathbf{w}$  satisfies one of the following conditions:

(1) supp(
$$\mathbf{w}$$
) = {(0,0), (d,0), (0,d), (i,d-i)} for some  $i = 1, ..., d-1$ ,

(2) 
$$supp(\mathbf{w}) = \{(0,0), (d,0), (0,d), (i,0)\} \text{ for some } i = 1, \dots, d-1,$$

(3) 
$$\operatorname{supp}(\mathbf{w}) = \{(0,0), (d,0), (0,d), (0,i)\} \text{ for some } i = 1, \dots, d-1.$$

Then, d = 2 and i = 1 hold.

*Proof.* Assume d > 2. Let **w** satisfy the third condition. Choose  $\lambda = (3, 1, ..., 1)$ . Then, apply Proposition 4.19. So, we have that the pairing matrix is invertible. So,  $\mathbf{w} = 0$  which is a contradiction. Thus, d = 2. By symmetry we have the same result for the second condition.

We want to show d = 2 for all outcomes **w** satisfying the first condition. Let **w**' satisfy the second condition. Then **w** = (123)**w**' holds. Assume d > 2. By Proposition 4.16, we have found an outcome **w**' of degree at least three. This contradicts Proposition 5.7 that we have just shown for the second condition. Thus, d = 2 holds.

Finally, we have 
$$i=1$$
 because  $i=1,\ldots,d-1$  and  $d=2$ .

**Proposition 5.8.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid integral outcome of degree d. Assume the outcome  $\mathbf{w}$  satisfies one of the following conditions:

(1) supp(
$$\mathbf{w}$$
) = {(0,0), (d,0), (0,e), (d-f,f)} for some e, f = 1,..., d-1.

(2) 
$$supp(\mathbf{w}) = \{(0,0), (0,d), (e,0), (d-f,f)\} \text{ for some } e, f = 1, \dots, d-1.$$

Then, d = 2 and e = f = 1 hold.

*Proof.* By Proposition 4.14, it suffices to show the statement for outcomes  $\mathbf{w}$  satisfying the first condition.

Let d > 2. If f = d - 1, then choose  $\lambda = (3, 1, ..., 1)$ . This allows us to apply Proposition 4.20 because  $0 + e \neq 2d - 1$  for d > 1. So,  $\mathbf{w} = \mathbf{0}$  which is a contradiction. Thus, we have f < d - 1. Then, we can choose  $\lambda = (2, 1, ..., 1)$ . Use Proposition 4.18 to get that  $\mathbf{w} = \mathbf{0}$ . This is a contradiction. Hence, we have d = 2.

Let 
$$d=2$$
. Then, we have  $e=f=1$  by definition of  $e$  and  $f$ .

Finally, we can prove Theorem 5.3.

Proof of Theorem 5.3. Use Proposition 5.5. For each case, either apply Proposition 5.6, Proposition 5.7, or Proposition 5.8 to show that the degree d equals two or three.

# Chapter 6

# Constraints on Supports of Valid Outcomes

We want to show in the next chapter that for all valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| = 4$  we have  $\deg(\mathbf{w}) \leq 2 \cdot |\operatorname{supp}^+(\mathbf{w})| - 3 = 5$ . For that goal we need new tools aside from the *Invertibility Criterion*. We will introduce the *Hyperfield Criterion* and *Contractions* to achieve it.

## 6.1 Hyperfield Criterion

In this section, we develop the *Hyperfield Criterion* which allows us to interpret Pascal forms as constraints on the support of valid outcomes.

Let us define the sign hyperfield. For some set A, the set  $2^A$  denotes the power set of A.

**Definition 6.1.** Let  $H := \{-1, 0, 1\}$ . We define the addition  $+: H \times H \to 2^H \setminus \{\emptyset\}$  on H as follows

$$0 + x = \{x\} \quad \forall x \in H, \quad 1 + 1 = \{1\}, \quad 1 + (-1) = H, \quad (-1) + (-1) = \{-1\}.$$

Multiplication  $\times : H \times H \to H$  is defined as usual. We call H the sign hyperfield.

Often, for singleton sets  $\{x\}$ , we will write x instead of  $\{x\}$ . So,

$$1+1=1$$
 or  $(-1)+0=-1$ .

**Remark 6.2.** The tuple  $(H, +, \cdot, 0, 1)$  is called a *hyperfield*. A hyperfield satisfies the following properties:

- (1) The maps + and  $\cdot$  are symmetric;
- (2)  $(H \setminus \{0\}, \cdot, 1)$  is a group;

- (3)  $0 \cdot x = 0$  and 0 + x = x hold for all  $x \in H$ ;
- (4)  $\bigcup_{q \in x+y} (q+z) = \bigcup_{q \in x+y} (x+q)$  hold for all  $x, y, z \in H$ ;
- (5)  $a \cdot (x+y) = (a \cdot x) + (a \cdot y)$  hold for all  $a, x, y \in H$ .
- (6) An inverse element  $y \in H$  exists for every  $x \in H$  such that the set x + y contains 0. This inverse element y is unique for every x and is denoted by -x.

Refer to [2] Section 6.1 or [1] for more details.

Next, we define polynomials over the sign hyperfield.

**Definition 6.3.** A polynomial in n variables  $x_1, \ldots, x_n$  over H is a formal sum

$$f = \sum_{\mathbf{k} \in \mathbb{Z}_{>0}^n} \lambda_{\mathbf{k}} \mathbf{x}^{\mathbf{k}}, \quad \lambda_{\mathbf{k}} \in H,$$

where only a finite number of coefficients  $\lambda_{\mathbf{k}}$  are non-zero, and  $\mathbf{x}^{\mathbf{k}} = x_1^{k_1} \cdots x_n^{k_n}$ . The set of all polynomials in n variables over H is denoted by  $H[x_1, \ldots, x_n]$ .

Let  $\mathbf{x} \in H$ . Then, we define

$$f(\mathbf{x}) \coloneqq \sum_{\mathbf{k} \in \mathbb{Z}_{>0}^n} \lambda_{\mathbf{k}} \mathbf{x}^{\mathbf{k}} \subset H.$$

We say that f vanishes at  $\mathbf{x} \in H$  if  $0 \in f(\mathbf{x})$ . In this case,  $\mathbf{x}$  is a hyperfield root of f.

Any real polynomial can be turned into a polynomial over the sign hyperfield by replacing the coefficients with elements of H. We can then evaluate the polynomial at any point in H.

**Definition 6.4.** Let  $f = \sum \lambda_{\mathbf{k}} \mathbf{x}^{\mathbf{k}} \in \mathbb{R}[\mathbf{x}]$  be a polynomial over  $\mathbb{R}$ . We call

$$\operatorname{sign}(f) := \sum_{\mathbf{k} \in \mathbb{Z}_{>0}^n} \operatorname{sign}(\lambda_{\mathbf{k}}) \mathbf{x}^{\mathbf{k}} \in H[\mathbf{x}]$$

the polynomial over H induced by f.

For sake of simplicity, we also write for any real vector  $\mathbf{w} \in \mathbb{R}^n$ :

$$\operatorname{sign}(\mathbf{w}) := (\operatorname{sign}(w_1), \dots, \operatorname{sign}(w_n)).$$

**Example 6.5.** Let d = 5. Consider the Pascal forms on  $\mathbb{Z}^{V_d}$  generated by diag(0), diag(1), diag(2), diag(3), diag(4) and diag(5). The polynomial over H induced by these forms can be depicted as follows:



**Definition 6.6.** A hyperfield Pascal form is just a polynomial over H induced by a Pascal form.

The reason we introduced the sign hyperfield is that it allows us to neglect the concrete values of the coefficients of a polynomial and focus on their signs. This makes reasoning about roots easier, which is helpful since we saw in earlier chapters that chipsplitting outcomes are roots of Pascal forms.

**Proposition 6.7.** Let  $f \in \mathbb{R}[\mathbf{x}]$  be a real polynomial. Let  $\mathbf{w} \in \mathbb{R}^n$  be a root of f. Then,  $\operatorname{sign}(\mathbf{w})$  is a root of  $\operatorname{sign}(f)$ .

*Proof.* Define  $\mathbf{s} := \operatorname{sign}(\mathbf{w})$ . Write  $f = \sum \lambda_{\mathbf{k}} \mathbf{x}^{\mathbf{k}}$  with real coefficients  $\lambda_{\mathbf{k}}$ . If  $\lambda_{\mathbf{k}} \mathbf{w}^{\mathbf{k}} = 0$  for all  $\mathbf{k} \in \mathbb{Z}_{\geq 0}^n$ , then clearly the sign of  $\lambda_{\mathbf{k}} \mathbf{w}^{\mathbf{k}}$  is zero; hence the sign of f is the singleton set  $\{0\}$  when evaluated at  $\mathbf{s}$ . So,  $\mathbf{s}$  is a root of  $\operatorname{sign}(f)$ .

Now, suppose that there exists some  $\mathbf{k} \in \mathbb{Z}_{\geq 0}^n$  such that  $\lambda_{\mathbf{k}} \mathbf{w}^{\mathbf{k}} \neq 0$ . Assume  $\lambda_{\mathbf{k}} \mathbf{w}^{\mathbf{k}} > 0$ . Then, there also exists some  $\mathbf{j} \in \mathbb{Z}_{\geq 0}^n$  such that we have  $\lambda_{\mathbf{j}} \mathbf{w}^{\mathbf{j}} < 0$ ; otherwise  $f(\mathbf{w}) > 0$  which

is a contradiction to **w** being a root of f. Thus,  $\operatorname{sign}(f)(\mathbf{s})$  has summands of both signs, and hence  $\operatorname{sign}(f)(\mathbf{s}) = H$ . So  $0 \in \operatorname{sign}(f)(\mathbf{s})$  holds. Therefore, **s** is a root of  $\operatorname{sign}(f)$ .

Taking the contrapositive of the above proposition, we get the *Hyperfield Criterion* which was first presented by Bik and Marigliano in [2].

**Proposition 6.8** (Hyperfield Criterion). Let  $\mathbf{s} = (s_{i,j})_{(i,j)\in V_d} \in H^{V_d}$ . Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a chipsplitting configuration with  $\operatorname{sign}(\mathbf{w}) = \mathbf{s}$ . If  $\mathbf{s}$  is not a root of a hyperfield Pascal form, then  $\mathbf{w}$  is not a chipsplitting outcome.

*Proof.* Follows from Proposition 6.7 and Theorem 3.13.

We call a vector  $\mathbf{s} \in H^{V_d}$  a sign configuration or hyperfield configuration. For completeness, we state standard definitions for sign configurations  $\mathbf{s} \in H^{V_d}$  similar to Definition 3.6.

**Definition 6.9.** Let  $\mathbf{s} \in H^{V_d}$  be a sign configuration. We define the following:

- (1) The positive support is defined as  $\operatorname{supp}^+(\mathbf{s}) := \{(i,j) \in V_d \mid s_{i,j} = 1\}.$
- (2) The negative support is defined as  $\operatorname{supp}^-(\mathbf{s}) := \{(i, j) \in V_d \mid s_{i,j} = -1\}.$
- (3) The support is defined as  $supp(s) := supp^+(s) \cup supp^-(s)$ .
- (4) The degree of **s** is defined as  $\deg(\mathbf{s}) := \max\{i + j \mid (i, j) \in \operatorname{supp}(\mathbf{s})\}.$
- (5) We call  $\mathbf{s}$  valid if its support is empty or supp<sup>-</sup>( $\mathbf{s}$ ) = {(0,0)}.

**Lemma 6.10.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a chipsplitting configuration. Then, the following statements hold:

- (1)  $\operatorname{supp}^+(\operatorname{sign}(\mathbf{w})) = \operatorname{sign}^+(\mathbf{w}),$
- $(2) \ \operatorname{supp}^-(\operatorname{sign}(\mathbf{w})) = \operatorname{sign}^-(\mathbf{w}),$
- (3)  $\deg(\operatorname{sign}(\mathbf{w})) = \deg(\mathbf{w}).$

*Proof.* Follows from the definitions.

To make use of the Hyperfield Criterion, we investigate hyperfield forms induced by the Pascal forms col(k), row(k), and diag(k).

**Proposition 6.11.** Let  $k = 0, \ldots, d$ . Define

$$A_k^+ := \{(i,j) \in V_d \mid j = 0, \dots, k \text{ and } i = k - j, \dots, d - j \text{ with } j \equiv k \pmod{2}\}$$

$$A_k^- := \{(i,j) \in V_d \mid j = 0, \dots, k \text{ and } i = k - j, \dots, d - j \text{ with } j \not\equiv k \pmod{2}\}$$

Then, the following statements hold:

(1) We have

$$\operatorname{sign}(\operatorname{diag}(k)) = \sum_{i=0}^{k} \sum_{j=0}^{d-k} x_{i,j}.$$

(2) We have

$$sign(col(k)) = \sum_{(i,j)\in A_k^+} x_{i,j} - \sum_{(i,j)\in A_k^-} x_{i,j}.$$

(3) We have

$$sign(row(k)) = \sum_{(i,j)\in A_k^+} x_{j,i} - \sum_{(i,j)\in A_k^-} x_{j,i}.$$

*Proof.* The first statement follows directly from Proposition 4.3 since  $i \leq k$  and  $d-i-j \geq k-i$  must hold for the binomial coefficient to be non-zero. The second and third statement follow similarly from Proposition 3.18.

**Proposition 6.12.** Let  $\mathbf{s} \in H^{V_d}$  be a valid nonzero sign configuration. The following statements hold:

- (1) Let k = 0, ..., d. If  $0 \in \text{sign}(\text{diag}(k))(\mathbf{s})$ , then  $\text{sign}(\text{diag}(k))(\mathbf{s}) = H$ .
- (2) If  $0 \in \text{sign}(\text{col}(k))(\mathbf{s})$  for all  $k = 0, \dots, d$ , then  $\text{sign}(\text{col}(k))(\mathbf{s}) = H$ .
- (3) If  $0 \in \text{sign}(\text{row}(k))(\mathbf{s})$  for all k = 0, ..., d, then  $\text{sign}(\text{row}(k))(\mathbf{s}) = H$ .

*Proof.* We see that **s** has at least degree  $d \ge 1$  since it is nonzero and valid. All  $s_{i,j}$  equal one for i+j>0, and there exists  $s_{k,d-k}=1$  for some  $k=0,\ldots,d$ .

(1) Assume sign(diag(k))(s) contains zero. By Proposition 6.11, we have

$$0 \in \operatorname{sign}(\operatorname{diag}(k))(\mathbf{s}) = \sum_{i=0}^{k} \sum_{j=0}^{d-k} s_{i,j}.$$

We know that  $s_{0,0}$  is minus one. So, we have  $s_{i,j} = 1$  for some i, j with i + j > 0. Thus, sign(diag(k))(s) = H.

(2) First note that  $\operatorname{col}(0) = \operatorname{diag}(d)$ . So, the case k = 0 is proven. Let k > 0. We start with k = d. Then, the union of  $A_d^+$  and  $A_d^-$  consists exactly of vertices of degree d. Since  $\operatorname{sign}(\operatorname{col}(d))(\mathbf{s}) = \sum_{(i,j) \in A_d^+} s_{i,j} - \sum_{(i,j) \in A_d^-} s_{i,j}$  contains zero, we have  $s_{i,j} = 1$  for some  $(i,j) \in A_d^+$ , and  $s_{i',j'} = -1$  for some  $(i',j') \in A_d^-$ . Hence,  $\operatorname{sign}(\operatorname{col}(d))(\mathbf{s}) = H$ .

Let k = d - 1. Then,  $s_{i,j} = 1$  for some  $(i,j) \in A_{k+1}^+$ , and  $s_{i',j'} = -1$  for some  $(i',j') \in A_{k+1}^-$ . Note that  $A_{k+1}^- \subset A_k^+$  by definition. Since  $\operatorname{sign}(\operatorname{col}(k))(\mathbf{s}) = \sum_{(i,j)\in A_k^+} s_{i,j} - \sum_{(i,j)\in A_k^-} s_{i,j}$  contains zero, we have  $s_{i'',j''} = -1$  for some  $(i'',j'') \in A_k^-$ . Hence,  $\operatorname{sign}(\operatorname{col}(k))(\mathbf{s}) = H$ .

Repeat this argument for k = d - 2, ..., 1 to show that sign(col(k))(s) = H.

(3) The proof is analogous to the previous case.

Corollary 6.13. Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid outcome. Then,

$$\operatorname{sign}(p)(\operatorname{sign}(\mathbf{w})) = H \quad \text{for all} \quad p \in \{\operatorname{diag}(k), \operatorname{col}(k), \operatorname{row}(k) \mid k = 0, \dots, d\}.$$

*Proof.* This follows from Theorem 3.13, Proposition 6.7, and Proposition 6.12.  $\Box$ 

**Example 6.14.** Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid outcome of degree d = 5. By the previous corollary and Example 6.5, we know that the outcome  $\mathbf{w}$  has at least one positive entry  $w_{i,j}$  in each of the following marked areas + because  $w_{0,0} < 0$ :

```
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Moreover, for each triangle below the outcome  $\mathbf{w}$  must have some  $w_{i,j} > 0$  for one vertex (i,j) in the plus area + and  $w_{i,j} > 0$  for another vertex (i',j') in the minus area - because  $\operatorname{sign}(\operatorname{col})(\operatorname{sign}(\mathbf{w})) = H$ .

Similarly, the statement holds for sign(row):



The above example demonstrates that we can view Corollary 6.13 as constraints on the support of a valid outcome  $\mathbf{w}$ . Configurations that do not satisfy these constraints are not valid outcomes.

Corollary 6.15. Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid outcome of degree  $d \geq 1$ . Then, all of the following constraints hold:

- (1) For all k = 0, ..., d, the positive support of **w** contains the vertex (i, j) for at least one i = 0, ..., k and j = 0, ..., d k.
- (2) For all k = 1, ..., d, the positive support of  $\mathbf{w}$  contains at least one  $(i, j) \in A_k^+$  and  $(i', j') \in A_k^-$ .
- (3) For all k = 1, ..., d, the positive support of  $\mathbf{w}$  contains at least one (i, j) and (i', j') from  $(j, i) \in A_k^+$  and  $(j', i') \in A_k^-$ .

We will use these constraints to efficiently compute all outcomes of degree some d.

## 6.2 Solving Hyperfield Linear Systems

We deviate from the approach by Bik and Marigliano in [2] by spending some considerate time in solving hyperfield linear systems. It will help us path the way for some breakthrough for the efficient computation of valid outcomes.

Fix some degree d. Due to the Hyperfield Criterion, we are interested in valid configurations  $\mathbf{w} \in \mathbb{Z}^{V_d}$  which satisfy

$$0 \in \operatorname{sign}(p)(\operatorname{sign}(\mathbf{w}))$$

for all  $p \in {\{\text{diag}(k), \text{row}(k), \text{col}(k)\}_{k=0}^d}$ . These configurations are potential valid outcomes. Let us consider how we can efficiently compute such valid configurations.

**Problem:** Given a set of linear forms  $A = \{p_1, \dots, p_k\}$ , compute the solution set  $V(A) := \{\mathbf{x} \in H^{V_d} : 0 \in \text{sign}(p_i)(\mathbf{x}) \quad \forall i = 1, \dots, k\}.$ 

We further simplify the problem by only considering configurations  $\mathbf{x}$  with supp<sup>-</sup>( $\mathbf{x}$ ) =  $\{(0,0)\}$  and  $|\operatorname{supp}^+(\mathbf{x})| = n$  for some fixed  $n \in \mathbb{N}$ .

**Problem:** Given a set of linear forms  $A = \{p_1, \dots, p_k\}$ , compute the solution set  $S_n(A) := V(A) \cap \{\mathbf{x} \in H^{V_d} : \operatorname{supp}^-(\mathbf{x}) = \{(0,0)\}, |\operatorname{supp}^+(\mathbf{x})| = n\}.$ 

Note that  $S_n(A)$  is a superset of valid outcomes of positive support size n, which will be useful in finding all valid outcomes.

### A Naive Approach

To compute  $S_n(A)$  a simple brute force algorithm can be used; just iterate over all positive support size n supports and check if they are hyperfield roots of some hyperfield Pascal basis.

#### **Algorithm 4** Brute Force Algorithm

```
Require: Positive support size n, a set of linear forms A = \{p_1, \ldots, p_k\}
Ensure: S_n(A)
 1: function Solve(A, n)
        initialize empty list solutions
 2:
        for n-combination S = \{(c_i, r_i) : i = 1, ..., n\} of V_d do
 3:
           initialize \mathbf{x} \in H^{V_d} with positive support S and x_{0,0} = -1
 4:
 5:
           if x is a hyperfield root of every p \in A then
               add S to solutions
 6:
           end if
 7:
        end for
 8:
        return solutions
 9:
10: end function
```

The naive approach has an exponential time complexity since  $\binom{(d+1)(d+2)/2}{n}$  many supports are checked.

## Efficient Algorithm

For a specific type of system of linear forms A, we can greatly speed up the computation of  $S_n(A)$ .

**Definition 6.16.** Let p be a linear form in  $H^{V_d}$ , and let  $\mathbf{x} \in H^{V_d}$  be some hyperfield root of p. If  $\text{supp}(\mathbf{x}) \cap \text{supp}(p) = \emptyset$ , then the root  $\mathbf{x}$  is called a *trivial root* of p. Otherwise, the root  $\mathbf{x}$  is called a *non-trivial root* of p.

**Definition 6.17.** Let A be a system of linear forms in  $H^{V_d}$ . We say  $S_n(A)$  is non-trivial if  $S_n(A) \neq \emptyset$  and every  $\mathbf{x} \in S_n(A)$  is a non-trivial root for every form  $p \in A$ . We say A is non-trivial if  $S_n(A)$  is non-trivial.

**Proposition 6.18.** Let A be a system of linear forms in  $H^{V_d}$ ,  $p \in A$  and  $\mathbf{x} \in S_n(A)$ . Then, the following statements hold:

```
(1) If (0,0) \in \text{supp}^+(p), then x_{i,j} = 1 for some (i,j) \in \text{supp}^+(p).
```

```
(2) If (0,0) \in \text{supp}^-(p), then x_{i,j} = 1 for some (i,j) \in \text{supp}^-(p).
```

Proof. Assume  $(0,0) \in \operatorname{supp}^+(p)$ . Since  $x_{0,0} = -1$ , we have  $-1 \in \operatorname{sign}(p)(x)$ . By assumption,  $\mathbf{x}$  is a hyperfield root of p, so  $0 \in \operatorname{sign}(p)(\mathbf{x})$ . This can only happen if  $x_{i,j} = 1$  for some  $(i,j) \in \operatorname{supp}^+(p)$ . The case  $(0,0) \in \operatorname{supp}^-(p)$  is similar.

The next proposition assumes that A is non-trivial.

**Proposition 6.19.** Let A be a non-trivial system of linear forms,  $p \in A$  and  $\mathbf{x} \in S_n(A)$ . If  $(0,0) \notin \operatorname{supp}(p)$ , then  $\operatorname{supp}^+(p) \neq \emptyset$ ,  $\operatorname{supp}^-(p) \neq \emptyset$  as well as  $x_{i,j} = x_{i',j'} = 1$  for some  $(i,j) \in \operatorname{supp}^+(p)$  and  $(i',j') \in \operatorname{supp}^-(p)$ .

*Proof.* Assume  $(0,0) \notin \text{supp}(p)$ . First,  $\text{supp}(p) \neq \emptyset$  because  $S_n(A)$  is non-empty and consists only of non-trivial roots. If  $\text{supp}^+(p) = \emptyset$ , then  $\text{supp}^+(x) \subset \text{supp}^-(p) = \text{supp}(p) \neq \emptyset$ . Hence,  $\text{sign}(p)(x) = \{-1\}$ , which contradicts  $\mathbf{x}$  being a root. Thus,  $\text{supp}^+(p)$  is non-empty. Similarly,  $\text{supp}^-(p)$  is non-empty.

By non-triviality,  $x_{i,j} = 1$  for some  $(i,j) \in \text{supp}(p)$ . Assume  $(i,j) \in \text{supp}^+(p)$ . Hence,  $1 \in \text{sign}(p)(\mathbf{x})$ . Since  $\mathbf{x}$  is a root, we also have  $0 \in \text{sign}(p)(\mathbf{x})$ . This can only occur if  $x_{i',j'} = 1$  for some  $(i',j') \in \text{supp}^-(p)$ . The case  $(i,j) \in \text{supp}^-(p)$  is similar.

Both propositions allow us to interpret hyperfield linear forms in a non-trivial system A as constraints on the positive supports of roots in  $S_n(A)$ .

**Example 6.20.** Fix the degree d = 3. Assume a system A and some linear form  $p \in A$ . Further assume p = diag(0) is a diagonal Pascal equation of order 0. The support of p is represented by the following diagram:

```
+
+ .
+ . .
```

We see that any nonzero hyperfield root  $\mathbf{x}$  of A satisfies  $x_i = 1$  for some

$$(i,j) \in \{(0,0), (0,1), (0,2), (0,3)\}$$

because  $x_{00}$  is negative.

Now, assume A is non-trivial. Consider a row pascal equation q = row(3) which is contained in A. Its support is depicted by the following diagram:

```
-
. +
. . -
. . . +
```

For some  $\mathbf{x}$  with supp<sup>-</sup>( $\mathbf{x}$ ) = {(0,0)} to be a hyperfield root of q, we must have either

- (1)  $x_{i,j} = x_{i',j'} = 1$  for some  $(i,j) \in \{(0,3),(2,1)\}$  and  $(i',j') \in \{(3,0),(1,2)\}$ , or
- (2)  $\operatorname{supp}(\mathbf{x}) \subset V_3 \setminus \operatorname{supp}(q)$ .

Considering only non-trivial roots  $\mathbf{x}$  lets us exclude the latter case. Thus, if we have a non-trivial system A with  $p, q \in A$ , to compute  $S_n(A)$ , it suffices to check only those hyperfield roots whose support intersected with each of the three following regions is non-empty:

Here are examples of such hyperfield roots illustrated in the following diagrams:

```
+
. +
. . . .
. . . .
```

**Definition 6.21.** To each linear form p in  $H^{V_d}$  we can associate a finite set of supports, which we call constraints $(p) := \{ \sup^+(p) \setminus \{(0,0)\}, \sup^-(p) \setminus \{(0,0)\} \} \subset 2^{V_d}$ .

The name is justified by the following proposition.

**Proposition 6.22.** Let p be a linear form in  $\mathbb{H}^{V_d}$ , and  $\mathbf{x} \in H^{V_d}$  with  $\operatorname{supp}^-(\mathbf{x}) = \{(0,0)\}$ . Then,  $\mathbf{x}$  is a non-trivial hyperfield root of p if and only if  $\operatorname{supp}^+(\mathbf{x}) \cap S \neq \emptyset$  for all  $S \in \operatorname{constraints}(p)$ .

*Proof.* Since  $\mathbf{x}$  is non-trivial, we clearly have non-empty intersection of  $\operatorname{supp}^+(\mathbf{x})$  and  $S \in \operatorname{constraints}(p)$ . The converse direction is also clear since  $p(\mathbf{x}) = 1 - 1 = H$ .

We present an algorithm for computing  $S_n(A)$  of non-trivial systems A.

#### **Algorithm 5** Algorithm for Non-Trivial Systems

**Require:** Positive support size n, non-trivial system A

Ensure:  $S_n(A)$ 

- 1: **function** SOLVE(A, n)
- 2:  $C \leftarrow \bigcup_{p \in A} constraints(p)$
- 3: solutions  $\leftarrow \{\mathbf{x} \in H^{V_d} \mid \forall S \in C : \operatorname{supp}^+(\mathbf{x}) \cap S \neq \emptyset, |\operatorname{supp}^+(\mathbf{x})| = n, \operatorname{supp}^-(\mathbf{x}) = \{(0,0)\}\}$
- 4: return solutions
- 5: end function

Proof of correctness. The correctness of solutions =  $S_n(A)$  follows from Proposition 6.22 and the assumption that A is non-trivial.

**Remark 6.23.** The algorithm can also be used for linear forms on  $H^{\Xi}$  with general index set  $\Xi \subset \mathbb{Z}_{\geq 0} \times \mathbb{Z}_{\geq 0}$ , and solutions  $\mathbf{x} \in H^{\Xi}$ . The proof is analogous.

## 6.3 Implementation of the Hyperfield Criterion

The Hyperfield Criterion states that only the common hyperfield roots of all Pascal forms can be supports of valid outcomes. The system of all Pascal forms is a-priori an *infinite* and *non-trivial* system. However, we found out that several bases of Pascal forms exist such as the row, col and diag Pascal basis. This allows us to consider *finite* systems. Define the finite system

$$A \coloneqq \{\operatorname{diag}(i)\}_{i=0}^d \cup \{\operatorname{row}(i)\}_{i=0}^d \cup \{\operatorname{col}(i)\}_{i=0}^d.$$

**Proposition 6.24.** The system A is non-trivial.

*Proof.* First,  $S_n(A)$  is non-empty because  $\mathbf{x} = (x_i)_{i=0}^n$  defined as  $x_{i,d-i} = \binom{n}{i}$  is a solution of the system A.

Let  $\mathbf{x} \in S_n(A)$  and  $i = 0, \dots, n$ . Consider the following cases.

- Assume,  $\mathbf{x} \notin \text{supp}(\text{diag}(i))$ ; then  $\text{diag}(i)(\mathbf{x}) < 0$ ; we found a contradiction to  $\mathbf{x}$  being a root.
- Assume,  $\mathbf{x} \notin \text{supp}(\text{row}(i))$ . If i = d, then  $\mathbf{x}$  is not of degree n. Therefore, we assume i < d. Then, either  $\mathbf{x}$  is a trivial root for row(i+1) or we have  $\text{row}(i+1)(\mathbf{x}) \neq 0$ . In the latter case, we found a contradiction to  $\mathbf{x}$  being a root. For the former case that  $\mathbf{x}$  is a trivial root, we conclude that there exists nonzero  $x_{u,d-u}$  for some  $u = i+2,\ldots,d$  since  $\mathbf{x}$  is of degree n; now we just repeat the argument for row(i+1). More precisely, if  $\mathbf{x}$  is again a trivial root for row(i+2), we repeat the argument for row(i+2) until we will end up with a contradiction  $\text{row}(u)(\mathbf{x}) \neq 0$ .

• For the case col(i), we can argue by symmetry.

**Corollary 6.25.** Configurations  $\mathbf{x} \in \mathbb{Z}^{V_d}$  of the form  $\operatorname{supp}(\mathbf{x}) \subset \{(i,j) \in \mathbb{Z}^{V_d} : i+j \leq k \text{ or } i > k+1\}$  are not valid outcomes for any  $k=0,\ldots,d-1$ . Neither are configurations  $\mathbf{x} \in \mathbb{Z}^{V_d}$  of the form  $\operatorname{supp}(\mathbf{x}) \subset \{(i,j) \in \mathbb{Z}^{V_d} : i+j \leq k \text{ or } j > k+1\}$  for  $k=0,\ldots,d-1$  due to symmetry.

*Proof.* Since the previously defined system A is non-trivial, we must have that supports of valid outcomes intersect the support of row(k+1) non-trivially.

**Example 6.26.** This is not a valid outcome:

Now that we have shown that A is a trivial system, we have found an efficient way to apply the Hyperfield Criterion. Here is a detailed breakdown of an implementation of the algorithm.

We implement the line of code  $C = \bigcup_{p \in A} \operatorname{constraints}(p)$  of Algorithm 5 using the procedure described in Algorithm 6. We can optimize the constraints by removing redundant constraints that are contained in each other, see Algorithm 7. To compute the solutions

$$\mathtt{solutions} = \{ \mathbf{x} \in H^{V_d} \mid \forall S \in C : \mathrm{supp}^+(\mathbf{x}) \cap S \neq \emptyset, |\mathrm{supp}^+(\mathbf{x})| = n, \mathrm{supp}^-(\mathbf{x}) = \{(0,0)\} \}$$

from Algorithm 5, we use the implementation detailed in Algorithm 8. To further enhance efficiency, we employ a heuristic that sorts the constraints by size and processes smaller constraints first (see line two of Algorithm 8). The complete implementation is written in Python 3 and included in the appendix TODO. Using this implementation, we can efficiently compute  $S_4(A)$ .

**Proposition 6.27.** Let  $A = \{\operatorname{diag}(i)\}_{i=0}^d \cup \{\operatorname{row}(i)\}_{i=0}^d \cup \{\operatorname{col}(i)\}_{i=0}^d \text{ for some degree } d \in \mathbb{N}.$  Let  $\mathbf{x} \in H^{V_d}$  be nonzero with  $\operatorname{supp}^-(\mathbf{x}) = \{(0,0)\}.$  The following statements hold:

(1) If d = 6, then we have  $\mathbf{x} \in S_4(A)$  if and only if  $\operatorname{supp}^+(\mathbf{x})$  is one of the following sets:

$$\{(0,3),(1,5),(4,1),(6,0)\},\{(0,5),(1,1),(3,3),(6,0)\},\{(0,6),(1,1),(3,3),(5,0)\},$$
  
 $\{(0,6),(1,1),(3,3),(6,0)\},\{(0,6),(1,4),(3,0),(5,1)\}.$ 

#### Algorithm 6 Make Constraints

```
Require: some hyperfield linear form p
Ensure: constraints(p)
 1: constraints ← list()
 2: for each (pos, neg) in support(p) do
        if 0 \in pos then
 3:
           new\_constr \leftarrow \{i \mid i \in pos \land i > 0\}
 4:
           if new\_constr \notin constraints then
 5:
               constraints.append(new_constr)
 6:
           end if
 7:
        else if 0 \in neq then
 8:
           new\_constr \leftarrow \{i \mid i \in neg \land i > 0\}
 9:
           if new\_constr \notin constraints then
10:
               constraints.append(new_constr)
11:
           end if
12:
        else if len(pos) > 0 \land len(neg) > 0 then
13:
           if pos \notin constraints then
14:
15:
               constraints.append(pos)
           end if
16:
           if neg \notin constraints then
17:
               constraints.append(neg)
18:
           end if
19:
20:
        end if
21: end for
22: return constraints
```

#### Algorithm 7 Remove Redundant Constraints

```
Require: constraints
Ensure: nonredundant constraints
 1: to_remove ← list()
 2: for each c \in \text{constraints do}
        for each d \in \text{constraints do}
 3:
            if c \supset d then
 4:
                to\_remove.append(c)
 5:
 6:
            end if
            if d \supset c then
 7:
                to\_remove.append(d)
 8:
            end if
 9:
        end for
10:
11: end for
12: return \{x \mid x \in \text{constraints} \land x \notin \text{to\_remove}\}
```

#### Algorithm 8 Solve

```
Require: positive support size n, non-trivial system A
Ensure: S_n(A)
 1: constraints ← remove_redundant_constraints(make_constraints(A))
 2: constraints.sort by length
 3: queue \leftarrow deque([()])
 4: for each constr \in constraints do
       for _ in range(|queue|) do
 5:
          conf ← queue.popleft()
 6:
          if conf intersects constr then
 7:
              queue.append(conf)
          else if |conf| < support_size then
 9:
              for each j \in \text{constr do}
10:
                 queue.append(conf \cup \{j\})
11:
              end for
12:
          end if
13:
       end for
14:
15: end for
16: return queue
```

- (2) If d = 7, then we have  $\mathbf{x} \in S_4(A)$  if and only if  $\operatorname{supp}^+(\mathbf{x})$  is one of the following sets:  $\{(0,7), (1,1), (3,3), (7,0)\}, \{(0,7), (1,3), (5,1), (7,0)\}, \{(0,7), (1,5), (3,1), (7,0)\}.$
- (3) If d = 8, 9, 10, 11, then the solution set  $S_4(A)$  is empty.

*Proof.* We compute the set of  $S_4(A)$  for d = 6, 7, 8, 9, 10, and 11 using the implementation of Algorithm 8 which is included in the appendix TODO. This gives us the results stated in the proposition.

We obtain the following corollary.

Corollary 6.28. For all valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| = 4$  and  $\deg(\mathbf{w}) \le 11$  we have  $\deg(\mathbf{w}) \le 5$ .

It remains to show the statement for cases  $deg(\mathbf{w}) > 11$ . We will do this by introducing contractions in the next section.

#### 6.4 Contractions

We aim to demonstrate that for every valid integral outcome  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| = 4$ , its degree is at most five. We have already shown this statement for outcomes of degree at

most eleven. A challenge arises because we need to check the set of valid outcomes of degree at least twelve, which is an infinite set. To address this, we introduce the technique of *contractions*. This approach allows us to reduce the problem to checking finitely many cases, while still accounting for all valid outcomes, even those with arbitrarily large degrees. For this, we build upon the method of *contractions* introduced by Bik and Marigliano in [2].

The idea of contraction is to *contract* or *consolidate* vertices in  $V_d$  by merging some rows and columns into a single vertex. We do this by defining new formal variables  $b_i$ ,  $c_i$ ,  $d_i$ ,  $e_i$ ,  $y_{i,j}$  and  $z_{i,j}$  called *contraction variables*.



Figure 6.1: This figure illustrates the contraction variables and is taken from [2]. The yellow areas X, Y, Z represent formal variables  $x_{i,j}$  that are unaffected by the contraction. The mint areas B, C, D represent rows and columns of vertices that are merged into a single vertex. The area D is further divided into areas  $D_1, D_2$  by alternating columns.

**Definition 6.29.** Let  $x_{i,j}$  be formal variables indexed by  $V_d$ . We merge a subset of rows and columns of formal variables  $x_{i,j}$  in  $V_d$  into a single vertex by defining the following contraction

variables:

$$y_{i,j} \coloneqq x_{i,d-3-i+j} \quad \text{for } i,j = 0, \dots, 3,$$

$$z_{i,j} \coloneqq x_{d-3-j+i,j} \quad \text{for } i,j = 0, \dots, 3,$$

$$b_j \coloneqq x_{4,j} + \dots + x_{d-4-j,j} \quad \text{for } j = 0, \dots, 3,$$

$$c_i \coloneqq x_{i,4} + \dots + x_{i,d-4-i} \quad \text{for } i = 0, \dots, 3,$$

$$d_k \coloneqq \begin{cases} x_{4,d-4-k} + x_{6,d-6-k} + \dots + x_{d-4-k,4} & \text{if } d+k \text{ is even} \\ x_{4,d-4-k} + x_{6,d-6-k} + \dots + x_{d-5-k,5} & \text{if } d+k \text{ is odd} \end{cases}$$

$$e_k \coloneqq \begin{cases} x_{5,d-5-k} + x_{7,d-7-k} + \dots + x_{d-5-k,5} & \text{if } d+k \text{ is even} \\ x_{5,d-5-k} + x_{7,d-7-k} + \dots + x_{d-4-k,4} & \text{if } d+k \text{ is odd} \end{cases}$$
for  $k = 0, \dots, 3$ .

Let us visualize the contraction variables for d = 16 in the following figure.

```
y_{0,3}
y_{0,2}
                    y_{1,3}
                   y_{1,2}
y_{0,1}
                                       y_{2,3}
                    y_{1,1}
                                        y_{2,2}
y_{0,0}
                                                             y_{3,3}
                    y_{1,0} y_{2,1} y_{3,2} d_0
   c_0
                                        y_{2,0} y_{3,1} d_1 e_0
   c_0
                       c_1
                                           c_2 y_{3,0} d_2 e_1 d_0
   c_0
                       c_1
                       c_1 \quad c_2 \quad c_3 \quad d_3 \quad e_2 \quad d_1 \quad e_0
   c_0
                       c_1 \quad c_2 \quad c_3 \quad * \quad e_3 \quad d_2 \quad e_1 \quad d_0
   c_0
                      c_0
   c_0
                      c_1 c_2 c_3 * * * * d_3 e_2 d_1
   c_0
   c_0
                   x_{1,3} x_{2,3} x_{3,3} b_3 b_3
x_{0,3}
                                       x_{2,2} x_{3,2} b_2 b_2 b_2 b_2 b_2 b_2 b_2 b_2 b_2 z_{0,2} z_{1,2} z_{2,2} z_{3,2}
x_{0,2}
                   x_{1.2}
                                                            x_{3,1} b_1 b_1 b_1 b_1 b_1 b_1
                                                                                                                                                                                                                  z_{0,1} z_{1,1} z_{2,1} z_{3,1}
                    x_{1.1}
                                        x_{2,1}
x_{0.1}
                   x_{1,0} x_{2,0} x_{3,0} b_0 b_0 b_0 b_0 b_0
                                                                                                                                                                                                                                      z_{0,0} z_{1,0} z_{2,0}
x_{0,0}
```

Figure 6.2: Contraction variables for d = 16 are depicted.

As we can see, the vertices  $x_{0,4}, \ldots, x_{0,d-4}$  are merged into the contraction variable  $c_0$  by summing them up. Similarly, the formal variables  $x_{i,4}, \ldots, x_{i,d-4-i}$  are merged into  $c_i$  for i = 1, 2, 3.

We remind that hyperfield Pascal forms are expressed as sums  $\sum_{(i,j)\in V_d} \lambda_{i,j} x_{i,j}$  with  $\lambda_{i,j} \in H$ . The key insight is that there are some Pascal forms that can be expressed in

terms of the contraction variables  $c_0, c_1, c_2$  and  $c_3$  instead of the original variables  $x_{i,j}$  for all vertices (i, j) in the C-area of Figure 6.4.

**Example 6.30.** Consider the Pascal form diag(1) in  $\mathbb{Z}^{V_{16}}$ . Its support is depicted in the following figure:

We see that diag(1) =  $x_{0,0} + x_{0,1} + x_{0,2} + x_{0,3} + x_{1,0} + x_{1,1} + x_{1,2} + x_{1,3} + y_{0,0} + y_{0,1} + y_{0,2} + y_{1,0} + y_{1,1} + y_{1,2} + y_{1,3} + c_0 + c_1$ .

The previous example also demonstrates that the expression diag(1) =  $x_{0,0} + x_{0,1} + x_{0,2} + x_{0,3} + x_{1,0} + x_{1,1} + x_{1,2} + x_{1,3} + y_{0,0} + y_{0,1} + y_{0,2} + y_{1,0} + y_{1,1} + y_{1,2} + y_{1,3} + c_0 + c_1$  is independent of the degree d, i.e. if we were to consider the Pascal form diag(1) in  $\mathbb{Z}^{V_d}$  for some arbitrary d, the expression would still hold. This is great news since it allows us to express Pascal forms in terms of contraction variables for all degrees d at once.

So far, we have only considered the contraction variables  $c_0$ ,  $c_1$ ,  $c_2$ , and  $c_3$ . As we might expect, we can also express some Pascal forms in terms of the contraction variables  $b_0$ ,  $b_1$ ,  $b_2$ ,  $b_3$ ,  $d_0$ ,  $d_1$ ,  $d_2$ ,  $d_3$ ,  $e_0$ ,  $e_1$ ,  $e_2$ ,  $e_3$ ,  $y_{i,j}$ , and  $z_{i,j}$ . We will now find these kinds of Pascal forms that can be represented in terms of the contraction variables independent of the degree d. A good set of Pascal forms to consider are the Pascal forms diag(k), col(k) and row(k) for k = 0, 1, 2, 3, d - 3, d - 2, d - 1, d.

**Proposition 6.31.** Let  $d \ge 11$ . Let p be a hyperfield form induced by one of the following Pascal forms on  $\mathbb{Z}^{V_d}$ :

- (2) row(1), row(2), row(3), or
- (3) diag(1), diag(2), diag(3), or
- (4) diag(d-1), diag(d-2), diag(d-3).

Then, we have

$$p = \sum_{i,j=0}^{3} \lambda_{i,j}^{(x)} x_{i,j} + \sum_{i,j=0}^{3} \lambda_{i,j}^{(y)} y_{i,j} + \sum_{i,j=0}^{3} \lambda_{i,j}^{(z)} z_{i,j} + \sum_{j=0}^{3} \lambda_{j}^{(b)} b_{j} + \sum_{i=0}^{3} \lambda_{i}^{(c)} c_{i} + \sum_{k=0}^{3} \lambda_{k}^{(d)} d_{k} + \sum_{k=0}^{3} \lambda_{k}^{(e)} e_{k}$$

with coefficients  $\lambda_{i,j}^{(x)}, \lambda_{i,j}^{(y)}, \lambda_{i,j}^{(z)}, \lambda_j^{(b)}, \lambda_i^{(c)}, \lambda_k^{(d)}, \lambda_k^{(e)} \in H$ .

Define  $\lambda(p,d) \coloneqq \left(\lambda_{i,j}^{(x)}, \lambda_{i,j}^{(y)}, \lambda_{i,j}^{(z)}, \lambda_j^{(b)}, \lambda_i^{(c)}, \lambda_k^{(d)}, \lambda_k^{(e)}\right)$  for the coefficients of p on  $H^{V_d}$ . Then, we have

$$\lambda(p,d) = \lambda(p,d+1) = \lambda(p,d+2) = \lambda(p,d+3) = \dots$$

In other words, all the coefficients  $\lambda_{i,j}^{(x)}, \lambda_{i,j}^{(y)}, \lambda_{i,j}^{(z)}, \lambda_{j}^{(b)}, \lambda_{i}^{(c)}, \lambda_{k}^{(d)}, \lambda_{k}^{(e)}$  are independent of the degree d.

*Proof.* For case two and three we observe that the hyperfield form p has support contained in the areas X, C, and Y from Figure 6.4. This follows directly from Proposition 6.11. We also see that p depends only on the column sums on C.

For case one and four we see that p has support contained in the areas X, B, and Z from Figure 6.4 by Proposition 6.11. We conclude that p depends only on the row sums on B.  $\square$ 

Here is another example.

**Example 6.32.** Consider diag(3) and d = 11.

Write

$$\operatorname{sign}(\operatorname{diag}(3)) = \sum_{i,j=0}^{3} x_{i,j} + \sum_{i=0}^{3} c_i + \sum_{i=0}^{3} \sum_{j=0}^{i} y_{i,j}.$$

This linear form is independent of the degree d.

**Proposition 6.33.** Let  $d \ge 12$ . Let p be a hyperfield form induced by one of the following Pascal forms on  $\mathbb{Z}^{V_d}$ :

- (1) col(d), col(d-1), col(d-2), col(d-3), or
- (2) row(d), row(d-1), row(d-2), row(d-3).

Then, we have

$$p = \sum_{i,j=0}^{3} \lambda_{i,j}^{(x)} x_{i,j} + \sum_{i,j=0}^{3} \lambda_{i,j}^{(y)} y_{i,j} + \sum_{i,j=0}^{3} \lambda_{i,j}^{(z)} z_{i,j} + \sum_{j=0}^{3} \lambda_{j}^{(b)} b_{j} + \sum_{i=0}^{3} \lambda_{i}^{(c)} c_{i} + \sum_{k=0}^{3} \lambda_{k}^{(d)} d_{k} + \sum_{k=0}^{3} \lambda_{k}^{(e)} e_{k}$$

with coefficients  $\lambda_{i,j}^{(x)}, \lambda_{i,j}^{(y)}, \lambda_{i,j}^{(z)}, \lambda_j^{(b)}, \lambda_i^{(c)}, \lambda_k^{(d)}, \lambda_k^{(e)} \in H$ .

Define  $\lambda(p,d) := \left(\lambda_{i,j}^{(x)}, \lambda_{i,j}^{(y)}, \lambda_{i,j}^{(z)}, \lambda_j^{(b)}, \lambda_i^{(c)}, \lambda_k^{(d)}, \lambda_k^{(e)}\right)$  for the coefficients of p on  $H^{V_d}$ . Then, we have

$$\lambda(p,d) = \lambda(p,d+2) = \lambda(p,d+4) = \lambda(p,d+6) = \dots$$
  
 $\lambda(p,d+1) = \lambda(p,d+3) = \lambda(p,d+5) = \lambda(p,d+7) = \dots$ 

In other words, all the coefficients of forms on even degrees are independent, and all the coefficients of forms on odd degrees are independent of the degree d.

*Proof.* All the hyperfield forms p depend on entries in the area Y, D, and Z from Figure 6.4 by Proposition 6.11. We see that p depends only on the alternating diagonal sums on D for  $x_{i,j}$  in the area D. This shows that p is a sum of the contraction variables  $y_{i,j}, z_{i,j}$ , and  $d_k - e_k$ .

**Example 6.34.** Consider col(d-3). Let d=12. The support of sign(col(d-3)) is depicted in the following figure:

We write

$$\operatorname{sign}(\operatorname{col}(d-3)) = \sum_{i=0}^{3} \sum_{j=0}^{i} (-1)^{i+j} y_{i,j} - \sum_{k=0}^{3} (-1)^k d_k + \sum_{k=0}^{3} (-1)^k e_k - \sum_{i,j=0}^{3} (-1)^j z_{i,j}.$$

For d = 13, the support of sign(col(d - 3)) is depicted in the following figure:

We write

$$\operatorname{sign}(\operatorname{col}(d-3)) = \sum_{i=0}^{3} \sum_{j=0}^{i} (-1)^{i+j} y_{i,j} - \sum_{k=0}^{3} (-1)^{k} d_{k} + \sum_{k=0}^{3} (-1)^{k} e_{k} + \sum_{i,j=0}^{3} (-1)^{j} z_{i,j}.$$

We have merged formal variables  $x_{i,j}$  indexed by vertices (i,j) in the areas B,C,D into contraction variables. Now, we apply these contractions to concrete elements  $\mathbf{s} \in H^{V_d}$ .

**Definition 6.35.** Define the index set

$$\Xi \coloneqq \{0,1,2,3\}^2 \sqcup \{0,1,2,3\}^2 \sqcup \{0,1,2,3\}^2 \sqcup \{0,1,2,3\} \sqcup \{0,1,2$$

Let  $\mathbf{s} \in H^{\Xi}$ . We call  $\mathbf{s}$  a contracted hyperfield configuration, and write  $\mathbf{s}$  as

$$\mathbf{s} = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}) = (x_{i,j}, y_{i,j}, z_{i,j}, b_j, c_i, d_k, e_k).$$

**Definition 6.36.** Let  $\mathbf{s} \in H^{\Xi}$  be a contracted hyperfield configuration. We say  $\mathbf{s}$  is *valid* if one of the following holds:

(1) s = 0 or

(2) 
$$s_{0,0} = -1$$
,  $x_{i,j} \ge 0$  for all  $i+j > 0$ , and  $y_{i,j}, z_{i,j}, b_j, c_i, d_k, e_k \ge 0$  for all  $i, j, k = 0, 1, 2, 3$ .

Going from the world of hyperfield configurations to the world of *contracted* hyperfield configurations is done via the following map.

**Definition 6.37.** Let  $d \ge 11$ . Let  $\mathbf{s} \in H^{V_d}$  be a hyperfield configuration. We define

$$\operatorname{contr}_d(\mathbf{s}) : \mathbf{s} \mapsto (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}) = (x_{i,i}, y_{i,i}, z_{i,i}, b_i, c_i, d_k, e_k)$$

where we set

$$x_{i,j} \coloneqq s_{i,j} \quad \text{for } i, j = 0, \dots, 3,$$

$$y_{i,j} \coloneqq s_{i,d-3-i+j} \quad \text{for } i, j = 0, \dots, 3,$$

$$z_{i,j} \coloneqq s_{d-3-j+i,j} \quad \text{for } i, j = 0, \dots, 3,$$

$$b_j \coloneqq s_{4,j} + \dots + s_{d-4-j,j} \quad \text{for } j = 0, \dots, 3,$$

$$c_i \coloneqq s_{i,4} + \dots + s_{i,d-4-i} \quad \text{for } i = 0, \dots, 3,$$

$$d_k \coloneqq \begin{cases} s_{4,d-4-k} + s_{6,d-6-k} + \dots + s_{d-4-k,4} & \text{if } d+k \text{ is even} \\ s_{4,d-4-k} + s_{6,d-6-k} + \dots + s_{d-5-k,5} & \text{if } d+k \text{ is odd} \end{cases}$$

$$e_k \coloneqq \begin{cases} s_{5,d-5-k} + s_{7,d-7-k} + \dots + s_{d-5-k,5} & \text{if } d+k \text{ is even} \\ s_{5,d-5-k} + s_{7,d-7-k} + \dots + s_{d-4-k,4} & \text{if } d+k \text{ is odd} \end{cases}$$
for  $k = 0, \dots, 3$ .

exactly as in Definition 6.29.

The contraction map contr<sub>d</sub> maps hyperfield configurations  $\mathbf{s} = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}) \in H^{V_d}$  to elements in  $H^{\Xi}$  if  $\mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e} \geq 0$ . If one of the entries is negative, the map may output to some element  $(2^H)^{\Xi}$ . To make life easier, we introduce the following definition.

**Definition 6.38.** Let  $\mathbf{s} \in H^{V_d}$  be a hyperfield configuration. We say  $\mathbf{s}$  is weakly valid if for all  $(i, j) \in \text{supp}^-(\mathbf{s})$  one of the following holds:

(1) 
$$i, j = 0, \dots, 3$$
, or

(2) 
$$i = 0, ..., 3$$
 and  $i + j > d - 3$ , or

(3) 
$$j = 0, ..., 3$$
 and  $i + j > d - 3$ .

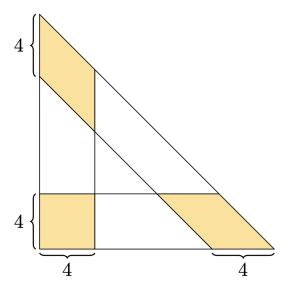


Figure 6.3: A hyperfield configuration is weakly valid if its negative support is contained in the yellow area. The figure is taken from [2].

From now on, we only consider weakly valid hyperfield configurations because in this case the contraction map contr<sub>d</sub> always outputs elements in  $H^{\Xi}$ .

**Definition 6.39.** Let  $\mathbf{s} \in H^{\Xi}$  be a contracted hyperfield configuration. The *positive support* of  $\mathbf{s} = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e})$  is defined as the set of all symbols  $x_{i,j}, y_{i,j}, z_{i,j}, b_j, c_i, d_k, e_k$  such that the corresponding coefficients of  $\mathbf{s}$  equal to one.

**Example 6.40.** Let  $\mathbf{s}=(\mathbf{x},\mathbf{y},\mathbf{z},\mathbf{b},\mathbf{c},\mathbf{d},\mathbf{e})\in H^\Xi$  be a contracted hyperfield configuration defined by

$$x_{0,0} = -1$$
,  $x_{0,3} = 1$ ,  $x_{1,1} = 1$ ,  $x_{3,0} = 1$ ,  $x_{0,0} = 1$ 

where all other entries are zero. Then, the positive support of s is given by

$$\operatorname{supp}^{+}(\mathbf{s}) = \{x_{0,3}, x_{1,1}, x_{3,0}, d_0, e_0\}.$$

# Chapter 7

# Valid Outcomes of Positive Support Size Four

We have all the tools to prove the main result of this chapter, namely for all valid integral outcomes  $\mathbf{w}$  with  $|\text{supp}^+(\mathbf{w})| = 4$  we have

$$\deg(\mathbf{w}) \leq 5.$$

As in previous chapters, we characterize outcomes as roots of Pascal forms. So, we define the following two systems of Pascal forms that valid outcomes must be roots of:

$$\Phi_1 := \{ \text{col}(1), \text{col}(2), \text{col}(3), \text{row}(1), \text{row}(2), \text{row}(3), \\ \text{diag}(1), \text{diag}(2), \text{diag}(3), \text{diag}(d-1), \text{diag}(d-2), \text{diag}(d-3) \},$$

and

$$\Phi_2 := \{ \operatorname{col}(d), \operatorname{col}(d-1), \operatorname{col}(d-2), \operatorname{col}(d-3), \\
\operatorname{row}(d), \operatorname{row}(d-1), \operatorname{row}(d-2), \operatorname{row}(d-3) \}.$$

We also define  $\Phi := \Phi_1 \cup \Phi_2$ .

By Proposition 6.31, we can write all hyperfield forms induced by Pascal forms p in  $\Phi_1$  as

$$\operatorname{sign}(p) = \hat{p}$$

for some linear form  $\hat{p} \in H[\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}]$  if  $d \geq 11$ . This linear form is independent of the degree d. To make notations consistent later, we set  $\hat{p}^{\text{even}} := \hat{p}^{\text{odd}} := \hat{p}$ .

Similarly, by Proposition 6.33, we can write all hyperfield forms induced by Pascal forms p in  $\Phi_2$  as

$$\operatorname{sign}(p) = \begin{cases} \hat{p}^{\text{even}} & \text{if } d \text{ is even} \\ \hat{p}^{\text{odd}} & \text{if } d \text{ is odd} \end{cases}$$

for some linear forms  $\hat{p}^{\text{even}}$ ,  $\hat{p}^{\text{odd}} \in H[\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}]$  if  $d \geq 12$ . These linear forms  $\hat{p}^{\text{even}}$ ,  $\hat{p}^{\text{odd}}$  are independent of the degree d.

**Definition 7.1.** We define the following three solution sets:

- (1) Define  $\Gamma_d$  to be the set of all valid hyperfield configurations  $\mathbf{s} \in H^{V_d}$  of degree d such that  $\operatorname{sign}(p)(\mathbf{s}) = H$  for all  $p \in \Phi$ .
- (2) Define  $\Gamma^{\text{even}}$  to be the set of all valid contracted hyperfield configurations  $\mathbf{s} \in H^{\Xi}$  such that  $\hat{p}^{\text{even}}(\mathbf{s}) = H$  for all  $p \in \Phi$ .
- (3) Define  $\Gamma^{\text{odd}}$  to be the set of all valid contracted hyperfield configurations  $\mathbf{s} \in H^{\Xi}$  such that  $\hat{p}^{\text{odd}}(\mathbf{s}) = H$  for all  $p \in \Phi$ .

By Proposition 6.8, valid chipsplitting outcomes of degree d have supports in  $\Gamma_d$ . This is the reason why we have defined  $\Gamma_d$  in the first place.

**Proposition 7.2.** Let  $d \ge 12$ . Then, the following holds:

- (1) If d is even, then  $\Gamma_d = \operatorname{contr}_d^{-1}(\Gamma^{\text{even}})$ .
- (2) If d is odd, then  $\Gamma_d = \operatorname{contr}_d^{-1}(\Gamma^{\operatorname{odd}})$ .

*Proof.* Let  $d \ge 12$  be even. Let  $\mathbf{s} \in H^{V_d}$  be a hyperfield configuration and  $p \in \Phi$ . Then, we have

$$\operatorname{sign}(p)(\mathbf{s}) = \hat{p}^{\operatorname{even}}(\operatorname{contr}_d(\mathbf{s})).$$

by definition of  $\hat{p}^{\text{even}}$ . If  $\mathbf{s} \in \Gamma_d$ , then  $H = \text{sign}(p)(\mathbf{s}) = \hat{p}^{\text{even}}(\text{contr}_d(\mathbf{s}))$ . Hence,  $\text{contr}_d(\mathbf{s})$  is contained in  $\Gamma^{\text{even}}$ . If  $\text{contr}_d(\mathbf{s}) \in \Gamma^{\text{even}}$  holds, using the equation above we also see that  $\mathbf{s} \in \Gamma_d$ . This shows that  $\Gamma_d = \text{contr}_d^{-1}(\Gamma^{\text{even}})$ .

The second statement for odd degrees d follows analogously.

Corollary 7.3. Let  $d \geq 12$ . Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid outcome. Then, we have

$$\operatorname{contr}_d(\operatorname{sign}(\mathbf{w})) \in \Gamma^{\operatorname{even}} \cup \Gamma^{\operatorname{odd}}.$$

*Proof.* Define  $\mathbf{s} := \operatorname{sign}(\mathbf{w})$ . By Proposition 6.12 we have  $\mathbf{s} \in \Gamma_d$ . If d is even, then  $\operatorname{contr}_d(\mathbf{s}) \in \Gamma^{\operatorname{even}}$  by the previous proposition. If d is odd, then  $\operatorname{contr}_d(\mathbf{s}) \in \Gamma^{\operatorname{odd}}$  by the previous proposition. This shows the claim.

This corollary allows us to exclude certain outcomes as valid outcomes. Assume we have some contracted hyperfield configuration  $\xi \in H^{\Xi}$  that is not a root of some of the linear forms  $\hat{p}^{\text{even}}, \hat{p}^{\text{odd}}$  for  $p \in \Phi$ . Then, any chipsplitting configuration  $\mathbf{w} \in \mathbb{Z}^{V_d}$  with  $\text{contr}_d(\text{sign}(\mathbf{w})) = \xi$  is not a valid outcome.

**Proposition 7.4.** Let  $\mathbf{s} \in H^{V_d}$  be a valid hyperfield configuration of degree d with positive support size four or less. If  $d \geq 12$ , then  $\mathbf{s} \notin \Gamma_d$ .

*Proof.* Let  $d \geq 12$ . For computing  $\Gamma_d$  we could use Algorithm 5 for all  $d = 12, 13, 14, \ldots$  and so on, which is not feasible since we would compute solutions sets for many infinitely many degrees d. Instead, we show that  $\Gamma^{\text{even}} \cup \Gamma^{\text{odd}}$  is empty. By Proposition 7.2,  $\Gamma_d$  is empty as well for all d > 12.

To show that  $\Gamma^{\text{even}}$  is empty, we can just use Algorithm 5 and Remark 6.23 with  $A := \{\hat{p}^{\text{even}} \mid p \in \Phi\}$ . Similarly, we compute  $\Gamma^{\text{odd}} = \emptyset$  with  $A := \{\hat{p}^{\text{odd}} \mid p \in \Phi\}$  and Algorithm 5. The results are in Appendix TODO. This shows the claim. TODO: show that A is non-trivial.

**Theorem 7.5.** For valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| = 4$  we have  $\deg(\mathbf{w}) \leq 5$ .

Proof. Let  $d \geq 6$ . Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid outcome with  $|\operatorname{supp}^+(\mathbf{w})| = 4$  and degree d. We have  $\operatorname{sign}(\mathbf{w}) \in \Gamma_d$ . By the previous proposition, there is no such  $\operatorname{sign}(\mathbf{w})$  for  $d \geq 12$ . By Proposition 6.27, the degree of  $\operatorname{sign}(\mathbf{w}) = d$  is six or seven. So, we just need to check eight cases. Of these eight cases, we can exclude all of them by applying Algorithm 2. The result of this algorithm is that only the zero outcome is possible for all these cases. This shows that the degree of  $\mathbf{w}$  is at most five.

# Chapter 8

# **Hexagon Criterion**

We now introduce the *Hexagon Criterion*, which allows us to determine if subconfigurations of a chipsplitting outcome are outcomes. It was first discovered in [2].

First, we need the following lemma to compute the determinant of matrix that we will encounter in the proof of the Hexagon Criterion.

**Lemma 8.1.** Let  $a, b, c \in \mathbb{Z}_{\geq 0}$ . Define the map  $H : \mathbb{Z}_{\geq 0} \to \mathbb{Z}_{\geq 0}, n \mapsto 0!1! \cdots (n-1)!$ ; note that H(0) = 1. Then, the following holds:

$$\det \left[ \binom{a+b}{b-i+j} \right]_{i,j=1}^{c} = \frac{H(a)H(b)H(c)H(a+b+c)}{H(b+c)H(c+a)H(a+b)}$$

Proof. See Theorem 8 of [4].

**Proposition 8.2** (Hexagon Criterion). Let  $d, d', \ell_1, \ell_2 \in \mathbb{N}$  with  $d' \geq 1$ ,  $\ell_1, \ell_2 \geq d'$ , and  $d' + \ell_1 + \ell_2 \leq d$ . Let  $\mathbf{w} = (w_{i,j})_{(i,j) \in V_d} \in \mathbb{Z}^{V_d}$  be a chip configuration. Define the subconfiguration  $\mathbf{w}' := (w_{i,j})_{(i,j) \in V_{d'}} \in \mathbb{Z}^{V_{d'}}$ . Assume the support of  $\mathbf{w}$  is not contained inside the "hexagon" (see Figure 8.1), or equivalently, assume that

$$supp(\mathbf{w}) \subset V_{d'} \cup \{(i,j) \in V_d \mid j > d - \ell_1\} \cup \{(i,j) \in V_d \mid i > d - \ell_2\}.$$

Then, the following holds:

- (1) If  $\mathbf{w}$  is an outcome, then also its subconfiguration  $\mathbf{w}'$  is an outcome.
- (2) If  $\mathbf{w}$  is a valid outcome, then  $\deg(\mathbf{w}) \leq d'$ .

Proof. We prove the first statement. Assume  $\mathbf{w}$  is an outcome. Let  $k = 0, \ldots, d'$  and  $\operatorname{diag}(\ell_1 + k) = \sum_{(i,j) \in V_d} \mu_{i,j} x_{i,j}$ . Consider the restricted linear form  $l_k = \sum_{(i,j) \in V_{d'}} \lambda_{i,j} x_{i,j}$ :  $\mathbb{Z}^{V_{d'}} \to \mathbb{Z}$  with  $\lambda_{i,j} := \mu_{i,j}$  for  $(i,j) \in V_{d'}$ . Then,  $l_0, \ldots, l_{d'}$  are Pascal forms on  $\mathbb{Z}^{V_{d'}}$  with  $l_k(\mathbf{w}') = \operatorname{diag}(\ell_1 + k)(\mathbf{w}) = 0$ . By Proposition 3.13 it suffices to show that  $l_0, \ldots, l_{d'}$  are linearly independent to show that  $\mathbf{w}'$  is an outcome.

Let a = 0, ..., d'. Define the configuration  $\mathbf{e}^{(a)} \in \mathbb{Z}^{V_{d'}}$  by

$$e_{i,j}^{(a)} := \begin{cases} 1 & \text{if } i = a \text{ and } j = d' - a, \\ 0 & \text{otherwise.} \end{cases}$$

Define the matrix

$$A := \left[ l_k(\mathbf{e}^{(a)}) \right]_{k,a=0}^{d'} = \left[ \binom{d-d'}{\ell_1 + k - a} \right]_{k,a=0}^{d'} = \left[ \binom{(d-d'-\ell_1) + \ell_1}{\ell_1 + k - a} \right]_{k,a=0}^{d'}.$$

We want to show that the matrix A is invertible because this implies that the linear forms  $l_0, \ldots, l_{d'}$  are linearly independent. We observe that

- (1) all entries of A are nonzero because  $0 \le \ell_1 + k a \le d d'$ , and
- (2)  $d d' \ell_1 \ge \ell_2 \ge 0$ .

This allows us to use Lemma 8.1 with

$$a := d - d' - \ell_1, \quad b := \ell_1, \quad c := d' + 1.$$

We obtain a nonzero determinant:

$$\det(A) = \frac{H(d - d' - \ell_1)H(\ell_1)H(d' + 1)H(d + 1)}{H(\ell_1 + d' + 1)H(1 + d - \ell_1)H(d - d')} \neq 0.$$

Hence,  $l_0, \ldots, l_{d'}$  are linearly independent.

For the second statement, let  $\mathbf{w}$  be an outcome, that is also valid. By the previous statement, the subconfiguration  $\mathbf{w}'$  is an outcome, as well. We extend  $\mathbf{w}'$  to some configuration  $\mathbf{v} \in \mathbb{Z}^{V_d}$  by

$$v_{i,j} := \begin{cases} w'_{i,j} & \text{if } (i,j) \in V_{d'}, \\ 0 & \text{otherwise.} \end{cases}$$

Clearly,  $\mathbf{v}$  is a *valid* outcome of degree at most d'. Then,  $\mathbf{v} - \mathbf{w}$  is an outcome with empty negative support. By Proposition 3.29,  $\mathbf{v} - \mathbf{w}$  is zero. Hence,  $\mathbf{w}$  has degree at most d'.  $\square$ 



Figure 8.1: This illustration is taken from [2]. If there is an outcome  $\mathbf{w}$  with support in the yellow area, then its subconfiguration  $\mathbf{w}'$  is an outcome if the support of  $\mathbf{w}'$  is contained in the bottom-left yellow triangle. Moreover, if the outcome  $\mathbf{w}$  is valid, then its support is only contained in the bottom-left yellow triangle.

# Chapter 9

# Valid Outcomes of Positive Support Size Five

In this chapter, we prove that for all valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| = 5$  we have

$$\deg(\mathbf{w}) \leq 7.$$

Tools like the Invertibility Criterion, Hyperfield Criterion, and the Hexagon Criterion will be used to prove this result. We follow the same proof as Bik and Marigliano in [2].

# **9.1** Case $d = 8, \dots, 41$

First, we show similar to Proposition 6.27 that no outcome of degree d = 8, ..., 41 exists with  $|\text{supp}^+(\mathbf{w})| = 5$ .

**Proposition 9.1.** Let  $A = \{\operatorname{diag}(i)\}_{i=0}^d \cup \{\operatorname{row}(i)\}_{i=0}^d \cup \{\operatorname{col}(i)\}_{i=0}^d \text{ for some degree } d \in \mathbb{N}.$ Let  $\mathbf{x} \in H^{V_d}$  be nonzero with  $\operatorname{supp}^-(\mathbf{x}) = \{(0,0)\}$ . Then, the number of solutions  $|S_5(A)|$  for  $d = 8, \ldots, 41$  is depicted in Table ??.

*Proof.* We compute the set of  $S_5(A)$  for d = 8, ..., 40, and 41 using the implementation of Algorithm 8 which is included in the appendix TODO.

**Proposition 9.2.** No outcome of degree d = 8, ..., 41 exists with  $|\text{supp}^+(\mathbf{w})| = 5$ .

*Proof.* We use Algorithm 2. The result of this algorithm is that only the zero outcome is possible for all these cases except for the following eight:

- $(1) \{(4,1),(5,0),(1,6),(0,8),(3,5)\},\$
- $(2) \ \{(3,0),(0,4),(5,1),(1,7),(4,4)\},$
- $(3) \{(3,0),(0,5),(5,1),(1,7),(4,4)\},\$

- $(4) \{(1,1),(3,0),(3,4),(0,8),(5,3)\},\$
- $(5) \{(3,0), (0,6), (5,1), (1,7), (4,4)\},\$
- (6)  $\{(0,3),(3,0),(5,1),(1,7),(4,4)\},\$
- $(7) \{(1,1),(4,0),(3,4),(0,8),(5,3)\},\$
- $(8) \{(3,1),(5,0),(1,6),(0,8),(3,5)\}.$

Valid configurations that have one of these eight sets as positive support are of degree eight or less.

Finally, we apply Algorithm 3 to these eight cases with  $E := \{3, 4, 5, 6, 7, 8\}$ . The result is that only the zero outcome is possible. This shows the claim.

## **9.2** Case d > 42

We have proven that no valid outcome of degree d = 8, ..., 41 exists with  $|\text{supp}^+(\mathbf{w})| = 5$ . Next, we show that no valid outcome of degree  $d \ge 42$  exists using contractions.

As in Proposition 7.4, we compute the sets

$$\Gamma^{\text{even}} \cap \left\{ \mathbf{s} \in H^{\Xi} : |\text{supp}^+(\mathbf{s})| \le 5 \right\} \text{ and } \Gamma^{\text{odd}} \cap \left\{ \mathbf{s} \in H^{\Xi} : |\text{supp}^+(\mathbf{s})| \le 5 \right\}$$

using Algorithm 5. The Proposition 7.4 has already shown that the sets

$$\Gamma^{\text{even}} \cap \left\{ \mathbf{s} \in H^{\Xi} : |\text{supp}^+(\mathbf{s})| \le 4 \right\} \text{ and } \Gamma^{\text{odd}} \cap \left\{ \mathbf{s} \in H^{\Xi} : |\text{supp}^+(\mathbf{s})| \le 4 \right\}$$

are empty. So, we just need to check the case  $|\text{supp}^+(\mathbf{s})| = 5$ .

**Definition 9.3.** We define

$$\Gamma_5^{\text{even}} \coloneqq \Gamma^{\text{even}} \cap \left\{ \mathbf{s} \in H^\Xi : |\text{supp}^+(\mathbf{s})| = 5 \right\} \text{ and } \Gamma_5^{\text{odd}} \coloneqq \Gamma^{\text{odd}} \cap \left\{ \mathbf{s} \in H^\Xi : |\text{supp}^+(\mathbf{s})| = 5 \right\}.$$

**Proposition 9.4.** The following holds:

- (1) We have  $|\Gamma_5^{\text{even}}| = 1283$ .
- (2) We have  $|\Gamma_5^{\text{odd}}| = 1265$ .

Proof. See APpendix TODO. # KEYWORD:LAOL

Corollary 9.5. Let  $d \ge 42$ . Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid outcome of degree d with  $|\operatorname{supp}^+(\mathbf{w})| = 5$ . Then,  $\operatorname{contr}_d(\operatorname{sign}(\mathbf{w})) \in \Gamma_5^{\operatorname{even}} \cup \Gamma_5^{\operatorname{odd}}$ .

*Proof.* This follows from Corollary 7.3.

Let us pick some  $\mathbf{s} \in \Gamma_5^{\text{even}} \cup \Gamma_5^{\text{odd}}$ . We want to show that any configuration  $\mathbf{w} \in \mathbb{Z}^{V_d}$  that maps to  $\mathbf{s}$  under  $\text{contr}_d \circ \text{sign}$  must be a zero outcome which is a contradiction. So, we need to check 1283 + 1265 = 2548 cases. Some of these cases occur multiple times; when we exclude these, we have 2318 cases to check:

$$|\Gamma_5^{\text{even}} \cup \Gamma_5^{\text{odd}}| = 2318.$$

**Remark 9.6.** If you compare this proof with the proof of Theorem 7.5, you will see that the number of cases in  $\Gamma^{\text{even}} \cup \Gamma^{\text{odd}}$  to check has increased from zero to 2318.

We make the following simplication to the index set  $\Xi$ .

**Definition 9.7.** Define the index set

$$\Xi' := \{0, 1, 2, 3\}^2 \sqcup \{0, 1, 2, 3\}^2 \sqcup \{0, 1, 2, 3\}^2 \sqcup \{0, 1, 2, 3\} \sqcup \{0, 1,$$

Let  $\mathbf{s} \in H^{\Xi}$  be weakly valid. We also define the map

$$\begin{split} \chi: H^\Xi &\to H^{\Xi'}, \\ \mathbf{s} &= (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}) \mapsto (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{\underline{d}} + \mathbf{\underline{e}}). \end{split}$$

The addition  $\mathbf{f} := \mathbf{d} + \mathbf{e}$  is done component-wise.

**Definition 9.8.** We define  $\operatorname{contr}_d := \chi \circ \operatorname{contr}_d$ .

**Definition 9.9.** Define  $\Lambda \subset H^{\Xi'}$  to be the set

$$\Lambda \coloneqq \chi(\Gamma_5^{\text{even}} \cup \Gamma_5^{\text{odd}}) = \left\{ \mathbf{s}' \in H^{\Xi'} \mid \mathbf{s}' = \chi(\mathbf{s}) \text{ for some } \mathbf{s} \in \Gamma_5^{\text{even}} \cup \Gamma_5^{\text{odd}} \right\}.$$

A simple computation of the set  $\Lambda$  shows that it has 2290 elements.

**Proposition 9.10.** Let  $s' \in \Lambda$ . Then, s' has positive support of size five or four.

*Proof.* Let  $\mathbf{s} = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}) \in \Gamma_5^{\text{even}} \cup \Gamma_5^{\text{odd}}$ . Since  $\mathbf{d}, \mathbf{e}$  are non-negative, we see from

$$\chi: H^{\Xi} \to H^{\Xi'}, \mathbf{s} \mapsto \mathbf{s}' = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d} + \mathbf{e})$$

that  $\mathbf{f} \in \{0, 1\}$  holds. Hence,  $\mathbf{s}'$  has positive support of size five or four.

```
y_{0,3}
y_{0,2}
       y_{1,3}
       y_{1,2}
y_{0,1}
               y_{2,3}
       y_{1,1}
               y_{2,2}
y_{0,0}
                       y_{3,3}
                              f_0
        y_{1,0}
                       y_{3,2}
 c_0
               y_{2,1}
                      y_{3,1} f_1 f_0
 c_0
        c_1
               y_{2,0}
                       y_{3,0} f_2 f_1 f_0
        c_1
                c_2
 c_0
                               f_3 f_2 f_1 f_0
                        c_3
 c_0
        c_1
                c_2
                                    f_3 f_2 f_1 f_0
                        c_3
 c_0
        c_1
                c_2
                                          f_3 f_2 f_1 f_0
        c_1
                        c_3
 c_0
                c_2
                                          * f_3 f_2 f_1
                                                                   f_0
 c_0
        c_1
                c_2
                        c_3
                                    *
                                                     f_3 f_2 f_1
                                                                          f_0
                        c_3
 c_0
        c_1
                c_2
                                                                                  f_0
        c_1
                c_2
                        c_3
                                                           f_3 f_2
                                                                          f_1
 c_0
                              b_3
                                   b_3 b_3 b_3 b_3
                                                           b_3 z_{0,3}
                                                                        z_{1,3} z_{2,3}
x_{0,3}
       x_{1,3}
               x_{2,3}
                      x_{3,3}
                                                                                        z_{3,3}
                                          b_2 b_2 b_2 b_2
                              b_2
                                    b_2
                                                                 b_2
       x_{1,2}
                                                                         z_{0,2} z_{1,2} z_{2,2}
x_{0,2}
               x_{2,2}
                       x_{3,2}
                                                                                               z_{3,2}
                               b_1
                       x_{3,1}
                                    b_1
                                          b_1
                                                b_1 b_1 b_1
                                                                   b_1
                                                                          b_1
x_{0,1}
       x_{1,1}
               x_{2,1}
                                                                                 z_{0,1} z_{1,1}
                                                                                                z_{2,1} z_{3,1}
                              b_0
                                         b_0
                                               b_0 b_0
                                                                                  b_0
                                    b_0
x_{0,0}
       x_{1,0}
              x_{2,0}
                      x_{3.0}
                                                                                        z_{0,0} z_{1,0} z_{2,0} z_{3,0}
```

Figure 9.1: Visualization of  $\Xi'$ .

## Algorithm 9 Check Configurations for Positive Support

9: return X

```
Ensure: a subset X \subset \Gamma_5^{\text{even}} \cup \Gamma_5^{\text{odd}} consisting of configurations that map to \mathbf{s}' \in \Lambda with positive support size four under \chi.

1: X \leftarrow \text{list}()
2: D \leftarrow \{d_0, d_1, d_2, d_3\}
3: E \leftarrow \{e_0, e_1, e_2, e_3\}
4: \mathbf{for} \ \mathbf{s} \in \Gamma_5^{\text{even}} \cup \Gamma_5^{\text{odd}} \ \mathbf{do}
5: \mathbf{if} \ \text{supp}(\mathbf{s}) \cap D \neq \emptyset \ \mathbf{and} \ \text{supp}(\mathbf{s}) \cap E \neq \emptyset \ \mathbf{then}
6: X \cdot \text{append}(\mathbf{s})
7: \mathbf{end} \ \mathbf{if}
8: \mathbf{end} \ \mathbf{for}
```

# 9.3 Case $d \ge 42$ Continued: $s' \in \Lambda$ with Positive Support Size Four

Corollary 9.11. From Proposition 9.10, we see that  $\mathbf{s}' \in \Lambda$  has positive support size four if and only if  $\mathbf{s} = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e})$  satisfies  $d_i = 1$  and  $e_j = 1$  for some i, j = 0, 1, 2, 3.

Corollary 9.12. Let  $\mathbf{s} \in \Gamma_5^{\text{even}} \cup \Gamma_5^{\text{odd}}$ . The element  $\mathbf{s}$  maps to some  $\mathbf{s}' \in \Lambda$  with positive support size four under  $\chi$  if and only if  $\text{supp}(\mathbf{s}) = \{x_{0,0}, x_{0,3}, x_{1,1}, x_{3,0}, d_0, e_0\}$ .

*Proof.* This is easily verified by computer using Algorithm 9. TODO TAG JAIKAJ.  $\Box$ 

We exclude this one case from the 2290 cases in  $\Lambda$  with the following proposition.

**Proposition 9.13.** Let  $d \geq 42$ . Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a weakly valid outcome. Then, we have

$$\operatorname{supp}^+(\operatorname{contr}_d(\operatorname{sign}(\mathbf{w}))) \neq \{x_{0,3}, x_{1,1}, x_{3,0}, d_0, e_0\}$$

*Proof.* Assume for the sake of contradiction that

$$supp^{+}(contr_{d}(sign(\mathbf{w}))) = \{x_{0.3}, x_{1.1}, x_{3.0}, d_{0}, e_{0}\}.$$

Then,  $\mathbf{w}$  is an outcome with support

$$supp(\mathbf{w}) = \{(0,0), (0,3), (1,1), (3,0), (i,d-i), (j,d-j)\}$$

for some even  $i = 4, 6, 8, \dots, d-4$  and odd  $j = 5, 7, 9, \dots, d-4$ .

Let  $\mathbf{u} \in \mathbb{Z}^{V_3}$  be the following outcome

It has support in supp( $\mathbf{u}$ ) =  $\{(0,0),(0,3),(1,1),(3,0)\}\subset \text{supp}(\mathbf{w})$ . Define the outcome  $\mathbf{v}:=\mathbf{w}+w_{0,0}\mathbf{u}$ . Then,  $v_{0,0}=0$  and  $\mathbf{v}\neq\mathbf{0}$ . However, if we apply the Invertibility Criterion with  $\lambda=\mathbf{1}$  on  $\mathbf{v}$ , we see that  $\mathbf{v}$  is zero. This is a contradiction. Hence,  $\text{supp}^+(\text{contr}_d(\text{sign}(\mathbf{w})))\neq\{x_{0,3},x_{1,1},x_{3,0},d_0,e_0\}$ .

# 9.4 Case $d \ge 42$ Continued: $s' \in \Lambda$ with Positive Support Size Five

It remains to show the other 2289 cases of  $\mathbf{s}' \in \Lambda$  with positive support size five. From now on, assume that  $\mathbf{s}' \in \Lambda$  has positive support size five. We will show that all these cases are invalid. For that we introduce *relative coordinates* defined below to make use of the Invertibility Criterion.

## Relative Coordinates and the Invertibility Criterion

**Definition 9.14.** Let  $d \ge 42$ . Let M be a sentinel value with no further significance. We use it to encode integers from  $4, \ldots, d-7$ . Define the map

$$\begin{aligned} \text{relcoord} : \{0, \dots, d\} &\to \{0, 1, 2, 3, d - 6, d - 5, d - 4, d - 3, d - 2, d - 1, d, M\} \,, \\ x &\mapsto \begin{cases} x & \text{if } x \in \{0, 1, 2, 3, d - 6, d - 5, d - 4, d - 3, d - 2, d - 1, d\} \,, \\ M & \text{if } x \in \{4, \dots, d - 7\} \,. \end{cases} \end{aligned}$$

Define the map relative set as follows:

relset : 
$$\mathbb{Z}^{V_d} \to 2^{\{0,\dots,3,M,d-6,\dots,d\} \times \{0,\dots,3,M,d-6,\dots,d\}}$$
,  
 $\mathbf{w} \mapsto \{(\text{relcoord}(i), \text{relcoord}(j)) \mid (i,j) \in \text{supp}(\mathbf{w})\}$ .

**Proposition 9.15.** Let  $d \ge 42$ . Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid outcome with positive support size five and degree d. Write  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{f})$ . Let i, j = 0, 1, 2, 3. Then, all of the following conditions hold:

- (1)  $(i, j) \in \text{relset}(\mathbf{w})$  if and only if  $x_{i,j} \neq 0$ ;
- (2)  $(i, d-3+j-i) \in \text{relset}(\mathbf{w})$  if and only if  $y_{i,j} \neq 0$ ;
- (3)  $(d-3+i-j,j) \in \text{relset}(\mathbf{w})$  if and only if  $z_{i,j} \neq 0$ ;
- (4) relset( $\mathbf{w}$ )  $\cap$  { $(M, i), (d 6, i), \dots, (d 4 i, i)$ }  $\neq \emptyset$  if and only if  $b_i \neq 0$ ;
- (5) relset( $\mathbf{w}$ )  $\cap \{(i, M), (i, d-6), \dots, (i, d-4-i)\} \neq \emptyset$  if and only if  $c_i \neq 0$ ;
- (6) relset(**w**)  $\cap \{(M, d-4-i), \dots, (M, d-6), (M, M), (d-6, M), \dots, (d-4-i, M)\} \neq \emptyset$  if and only if  $f_i \neq 0$ .

*Proof.* Let **w** be some valid outcome with  $x_{i,j} \neq 0$ . Then,  $w_{i,j} \neq 0$  with i, j = 0, 1, 2, 3. By definition of relcoord,  $i \mapsto i$  and  $j \mapsto j$ . So  $(i,j) \in \text{relset}(\mathbf{w})$  since  $(i,j) \in \text{supp}(\mathbf{w})$ .

Assume  $y_{i,j} \neq 0$ . Then,  $w_{i,d-3+j-i} \neq 0$ . By definition of relcoord,  $i \mapsto i$  and  $d-3+j-i \mapsto d-3+j-i$ . So  $(i,d-3+j-i) \in \text{relset}(\mathbf{w})$  since  $(i,d-3+j-i) \in \text{supp}(\mathbf{w})$ . The case  $z_{i,j} \neq 0$  is similar.

Assume  $b_i \neq 0$ . Then, there must exist some nonzero  $w_{k,i}$  for some  $k = 4, \ldots, d - 4 - i$ . Clearly, k maps to some element in  $\{M, d - 6, \ldots, d - 4 - i\}$ . This shows the claim. The case for  $c_i$  is similar.

Assume  $f_i \neq 0$ . Then, there must exist some nonzero  $w_{k,d-i-k}$  for some  $k = 4, \ldots, d-4-i$ . Clearly, k and d-i-k map to some element in  $\{M, d-6, \ldots, d-4-i\}$ . This shows the claim.

**Remark 9.16.** Given some  $\mathbf{s}' \in H^{\Xi'}$  we can compute all the relative supports relset( $\mathbf{w}$ ) for all  $\mathbf{w} \in \mathbb{Z}^{V_d}$  with  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = \mathbf{s}'$ . Proposition 9.15 gives us an easy way to compute this set of relative supports by interpreting all the six conditions 1., 2., 3., 4., 5., and 6. in Proposition 9.15 as constraints on the relative support of  $\mathbf{w}$ .

### Algorithm 10 Make relative constraints

Require: some nonzero coordinate  $t \in \{x_{i,j}, y_{i,j}, z_{i,j}, b_i, c_i, f_i\}$  of  $\mathbf{s}' = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{f}) \in H^{\Xi'}$ Ensure: set of constraints on the relative support of  $\mathbf{w}$  that correspond to the nonzero coordinate t as in Proposition 9.15.

```
1: function MAKE_REL_CONSTRAINTS(t)
       if t = x_{i,j} then
2:
           return [(i,j)]
3:
       end if
4:
       if t = y_{i,j} then
5:
           return [(i, d - 3 + j - i)]
6:
       end if
7:
       if t = z_{i,j} then
8:
           return [(d-3+i-j,j)]
9:
       end if
10:
       if t = b_i then
11:
           return \{(M, i), (d - 6, i), \dots, (d - 4 - i, i)\}
12:
       end if
13:
14:
       if t = c_i then
           return \{(i, M), (i, d-6), \dots, (i, d-4-i)\}
15:
       end if
16:
       if t = f_i then
17:
           return \{(M, d-4-i), \dots, (M, d-6), (M, M), (d-6, M), \dots, (d-4-i, M)\}
18:
19:
       end if
20: end function
```

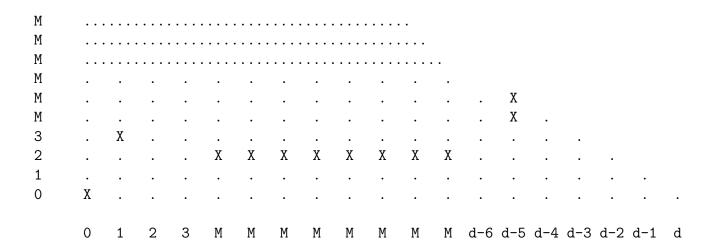
Relative coordinates help us to apply the Invertibility Criterion.

**Example 9.17.** Let  $d \geq 42$ . Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be some valid configuration with support size six. Now, assume that the relative support set of  $\mathbf{w}$  is

relset(
$$\mathbf{w}$$
) = {(0,0), (0, d), (1,3), (M,2), (M, d-6), (d-5, M)}.

Let us visualize the configuration w. Can such a configuration exist?

```
d X
d-1 . .
d-2 . . .
d-3 . . . .
d-4 . . . . .
d-5 . . . . . .
X X
M . . . . . . . . . .
```



We see that

$$\operatorname{supp}(\mathbf{w}) = \{(0,0), (0,d), (1,3)\} \cup \{(i,2), (j,d-6)\} \cup \{(d-5,k)\}.$$

for i, j, k = 4, ..., d - 7. When i = j, we can apply the Invertibility Criterion (Proposition 4.20 and Proposition 4.18) with  $\lambda = (3, 1, ..., 1, 2, 1, ..., 1)$  to see that  $\mathbf{w} = \mathbf{0}$ , which is a contradiction. So assume  $i \neq j$ . Then, we use  $\lambda = (3, 1, ..., 1)$  to see that  $\mathbf{w} = \mathbf{0}$ , which is a contradiction. Hence,  $\mathbf{w}$  cannot be an outcome.

Let us generalize this example to elements  $\mathbf{s}' \in \Lambda$ .

**Proposition 9.18.** Let  $\mathbf{s}' \in \Lambda$  with positive support size five. Then, all of the following hold:

- (1)  $|\operatorname{supp}^+(\mathbf{s}') \cap \{b_0, b_1, b_2, b_3\}| \le 1;$
- (2)  $|\operatorname{supp}^+(\mathbf{s}') \cap \{c_0, c_1, c_2, c_3\}| \le 1;$
- (3)  $|\operatorname{supp}^+(\mathbf{s}') \cap \{f_0, f_1, f_2, f_3\}| \le 1.$

*Proof.* This is verified by computer. Keyword TAG MUSIAS.

Corollary 9.19. Let  $\mathbf{s}' \in \Lambda$  with positive support size five. Then, we have

$$|\{(x,y)\in \operatorname{relset}(\mathbf{s}'): x=M\}| \leq 2 \quad and \quad |\{(x,y)\in \operatorname{relset}(\mathbf{s}'): y=M\}| \leq 2.$$

*Proof.* Follows immediately from the previous proposition.

**Proposition 9.20.** Let  $d \geq 42$ . Let  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be a valid configuration with positive support size five and

$$|\{(x,y) \in \text{relset}(\mathbf{w}) : x = M\}| = 2.$$

Denote these elements by (M, x) and (M, y) for  $x \neq y$ . Write  $(i, x), (i', y) \in \text{supp}(\mathbf{w})$  with  $i, i' = 4, \ldots, d-7$  for the elements that map to (M, x) or (M, y) under relcoord, respectively. If we successfully apply the Invertibility Criterion for the case i = i' to show the contradiction  $\mathbf{w} = \mathbf{0}$  with

$$\lambda = (\mathbf{a}, 1, \dots, 1, 2, 1, \dots, 1, \mathbf{b})$$

for some  $\mathbf{a} \in \mathbb{Z}^k_{\geq 1}$ ,  $\mathbf{b} \in \mathbb{Z}^h_{\geq 1}$ , and  $1 \leq k, h \leq 4$ , then we can also apply the Invertibility Criterion for the case  $i \neq i'$  with

$$\lambda' = (\mathbf{a}, 1, \dots, 1, \mathbf{b})$$

to show the same contradiction.

*Proof.* Assume  $i \neq i'$ . Then, we can just apply Proposition 4.17 as long as  $S'_{l'} \in \{0, \lambda'_l\}$  is satisfied for all l', see Section 4.2 on the divide and conquer approach of the Invertibility Criterion. Theses conditions are satisfied because the sets  $S_l$  induced by  $\lambda$  satisfy  $S_l \in \{0, \lambda_l\}$  by assumption.

**Corollary 9.21.** The previous propostion allows us to specialize to the case i = i' when  $(i, x), (i', y) \in \text{supp}(\mathbf{w})$  for i, i' = 4, ..., d - 7 and  $x \neq y$  occurs. A similar statement holds for the case  $|\{(x, y) \in \text{relset}(\mathbf{w}) : y = M\}| = 2$ .

**Example 9.22.** Returning to Example 9.17, we see that it suffices to consider the case i = j. The case  $i \neq j$  then follows from the previous corollary.

Corollary 9.23. Let  $d \geq 42$ . Let  $S = \{(i, x), (i, y), (i + 1, z)\} \subset V_d$ . If  $\operatorname{relcoord}(x) \in \{0, 1, 2, 3\}$ ,  $\operatorname{relcoord}(y) \in \{d - 6, \dots, d\}$  and  $\operatorname{relcoord}(z) \neq M$ , then  $A_{E,S}^{(d)}$  is invertible.

*Proof.* Use Proposition 4.20. We just need to make sure that  $x+y\neq 2z+1$ . Assume x+y=2z+1. Then,  $z=\frac{x+y-1}{2}$ . The smallest value for z is  $\frac{d-6-1}{2}=17.5$ , and the largest value is  $\frac{d+3-1}{2}=22$  for d=42. Hence, z does map to M under relative coordinates for all d=42 and for d>42 we well. By assumption, relcoord $(z)\neq M$ . This is a contradiction, so  $x+y\neq 2z+1$  for all  $d\geq 42$ .

The corollary above can be extended to several other cases which allows us to implement an algorithm to exclude certain cases of  $\mathbf{s}' \in \Lambda$ . First, we need an algorithm for the division step.

This division algorithm just implements the division rule to choose  $\lambda_i$  minimal such that  $|S_i| \in \{0, \lambda_i\}$  as explained in Remark 4.8.

**Example 9.24.** Here are some examples of the algorithm in action.

- (1) L = [3, 0, 2, 0, 0, 0, 0, 0, 0, 1] for support [(0, 0), (0, 3), (1, d 1), (M, 1), (M, M), (d, 0)].
- $(2) \ \ L = [3,0,2,0,0,0,0,0,0,1] \ \text{for support} \ [(0,0),(0,3),(1,d-1),(M,1),(M,d-6),(d,0)].$

#### Algorithm 11 Divide

```
Require: relset: Support of a configuration in relative coordinates
Ensure: L \in \mathbb{Z}^k where L_i = |S_i| if a division (\lambda, (E_i)_{i=1}^k, (S_i)_{i=1}^k) is found; None otherwise
 1: M \leftarrow 500, d \leftarrow 1000
 2: R \leftarrow [0, 1, 2, 3, M, d - 6, d - 5, d - 4, d - 3, d - 2, d - 1, d]
 3: L \leftarrow list(), col_start \leftarrow 0
 4: for col_end \in \{0, \dots, |R| - 1\} do
        num\_cols \leftarrow col\_end - col\_start + 1
        points_in_col \leftarrow \{p = (x, y) \in relset \mid x \in [R[col_start], R[col_end]]\}
 6:
        num_points \leftarrow |points_in_col|
 7:
        if (num\_points = 0) \lor (num\_cols = num\_points) \lor (R[col\_end] = M) then
 8:
             L.append(num\_points)
 9:
            col\_start \leftarrow col\_end + 1
10:
11:
        end if
12: end for
13: if sum(L) \neq 6 then
        return None
15: end if
16: \mathbf{return}\ L
```

(3) L = [3, 0, 0, 0, 2, 0, 0, 0, 1] for support [(0, 0), (0, 3), (1, d-1), (d, 0), (d-5, 1), (d-5, M)].

**Proposition 9.25.** Let  $d \ge 42$ . Let  $a, b = 0, \ldots, d$  with  $b \ge a$  Then, the following hold:

(1) If 
$$b = 0, 1, 2, 3$$
 or  $a = d - 6, \dots, d$ , then  $relcoord(\frac{a+b-1}{2}) \neq M$ .

(2) Let  $b = 4, \ldots, d - 7$ .

a) If 
$$a = 0, 1$$
, then relcoord $(\frac{a+b-1}{2}) \in \{2, 3, M\}$ .

b) If 
$$a = 2, 3$$
, then  $\operatorname{relcoord}(\frac{a+b-1}{2}) \in \{3, M\}$ .

(3) Let  $a = 4, \ldots, d - 7$ .

a) If 
$$b = d - 6, d - 5$$
, then relcoord $(\frac{a+b-1}{2}) = M$ .

b) If 
$$b = d - 4, d - 3$$
, then  $relcoord(\frac{a+b-1}{2}) \in \{M, d - 6\}$ .

c) If 
$$b = d - 2, d - 1$$
, then relcoord $(\frac{a+b-1}{2}) \in \{M, d - 6, d - 5\}$ .

d) If 
$$b = d$$
, then relcoord $(\frac{a+b-1}{2}) \in \{M, d-6, d-5, d-4\}$ .

*Proof.* Let  $d \geq 42$ .

(1) Let a=b=3. Then  $\frac{6-1}{2}=2.5$ . Thus, for all  $a\leq b\leq 3$  we have  $\frac{a+b-1}{2}\leq 2.5$ , so  $\operatorname{relcoord}(\frac{a+b-1}{2})\neq M$ . Let a=b=d-6. Then  $\frac{a+b-1}{2}>d-7$ . Thus, for all  $d-6\leq a\leq b\leq d$  we have  $\operatorname{relcoord}(\frac{a+b-1}{2})\neq M$ .

- (2) Let  $b = 4, \ldots, d 7$ .
  - a) For a=0 and b=4 we obtain  $\frac{a+b-1}{2}>1.5$ . So,  $\operatorname{relcoord}(\frac{a+b-1}{2})\notin\{0,1\}$ . For a=1 and b=d-7, we see that  $\frac{a+b-1}{2}=\frac{d-7}{2}$ , which maps to M under relative coordinates for  $d\geq 42$ . So the midpoint maps to values between two and M under relcoord, i.e.  $\{2,3,M\}$ . This shows the claim.
  - b) The proof of case a = 2, 3 is similar to the previous proof.
- (3) Let  $a = 4, \ldots, d 7$ .
  - a) Let b=d-6, d-5. Then,  $\frac{a+b-1}{2} \geq \frac{d-3}{2} \geq 19.5$  for all  $d \geq 42$ . This maps to M under relative coordinates. Moreover,  $\frac{a+b-1}{2} \leq \frac{d-7+d-5-1}{2} = \frac{2d-13}{2} = d-6.5$ . Thus,  $\operatorname{relcoord}(\frac{a+b-1}{2}) = M$ .
  - b) The proof of cases  $b = d 4, d 3, \dots, d$  are similar to the previous proof.

This proposition gives Algorithm 12 to compute the midpoint of two points in relative coordinates. Next, we define the order in relative coordinates:

$$0 < 1 < 2 < 3 < M < d - 6 < d - 5 < d - 4 < d - 3 < d - 2 < d - 1 < d$$

The **succ** function is defined as the successor in this order, see Algorithm 13. We can now implement the conquer step of the divide and conquer approach of the Invertibility Criterion, see Algorithm 14. This leads us to an implementation of the Invertibility Criterion for relative coordinates, see Algorithm 15.

Let us apply the Invertibility Criterion for relative coordinates to all of the 2289 cases of  $\mathbf{s}' \in \Lambda$  with positive support size five. Let  $\mathbf{s}'$  be one of these cases. We need to compute the set

$$R(\mathbf{s}') := \{ \text{relset}(\mathbf{w}) \mid \mathbf{w} \in \mathbb{Z}^{V_d} \text{ such that } \text{contr}'(\text{sign}(\mathbf{w})) = \mathbf{s}' \}.$$

Here is a simple depth first search algorithm to compute  $R(\mathbf{s}')$  using Proposition 9.15, see Algorithm 17. Now, we can apply the Invertibility Criterion for relative coordinates to all of  $\mathbf{m} \in R(\mathbf{s}')$ . If all of these applications to  $\mathbf{m} \in R(\mathbf{s}')$  are successful, then  $\mathbf{s}'$  is impossible. Otherwise, we cannot exclude  $\mathbf{s}'$  as a possible outcome.

Of the 2289 cases of  $\mathbf{s}' \in \Lambda$  with positive support size five, we rule out 1182 cases by applying the Invertibility Criterion for relative coordinates using Algorithm 18. The remaining 1107 cases are inconclusive. The details of the implementation can be found in the Appendix TODO.

## **Algorithm 12** Compute midpoint $\frac{a+b-1}{2}$

```
Require: Constants: SENTINEL_M \leftarrow 500, SENTINEL_D \leftarrow 1000
 1: function midpoint(a, b)
       low, high \leftarrow \min(a, b), \max(a, b)
 2:
       if low > SENTINEL_M or high < SENTINEL_M then
 3:
           return \left\{\frac{low+high-1}{2}\right\}
 4:
       else if high = SENTINEL_M then
 5:
           if low \in \{0,1\} then
 6:
              return {2,3,SENTINEL_M}
 7:
           else if low \in \{2,3\} then
 8:
              return {3, SENTINEL_M}
 9:
           end if
10:
       else if low = SENTINEL_M then
11:
           if high \in \{SENTINEL_D - 6, SENTINEL_D - 5\} then
12:
              return {SENTINEL_M}
13:
           else if high \in \{SENTINEL_D - 4, SENTINEL_D - 3\} then
14:
              return {SENTINEL_M, SENTINEL_D -6}
15:
           else if high \in \{SENTINEL_D - 2, SENTINEL_D - 1\} then
16:
              return {SENTINEL_M, SENTINEL_D -6, SENTINEL_D -5}
17:
           else if high = SENTINEL_D then
18:
              return {SENTINEL_M, SENTINEL_D -6, SENTINEL_D -5, SENTINEL_D -4}
19:
           end if
20:
21:
       end if
       return {SENTINEL_M}
22:
23: end function
```

#### Algorithm 13 Successor

```
Require: Constants: SENTINEL_M \leftarrow 500, SENTINEL_D \leftarrow 1000
 1: function succ(x)
       if x = 3 then
 2:
 3:
          return SENTINEL_M
 4:
       else if x = SENTINEL_M then
          return SENTINEL_D -6
 5:
       else
 6:
          return x+1
 7:
       end if
 9: end function
```

## Algorithm 14 Conquer Algorithm

Require: support relset in relative coordinates

```
Ensure: True only if the Invertibility Criterion is successful; False if inconclusive
 1: function conquer(relset)
        relset ← relset.sort_by_column()
        length \leftarrow |relset|
 3:
        if length \leq 2 then
 4:
            return True
                                                                        \triangleright Proposition 4.17 and 4.18
 5:
        else if length = 3 then
 6:
            x, y, z \leftarrow \texttt{relset}[0], \texttt{relset}[1], \texttt{relset}[2]
 7:
            same\_column \leftarrow (col(x) = col(y) = col(z))
 8:
            if same_column then
                                                                                  \triangleright Proposition 4.19
 9:
                return True
10:
            else if col(x) = col(y) and succ(col(x)) = col(z) then
11:
               if row(z) \notin midpoint(row(x), row(y)) then
                                                                                  \triangleright Proposition 4.20
12:
                   return True
13:
               end if
14:
            end if
15:
            return False
16:
17:
        else
            return False
18:
19:
        end if
20: end function
```

#### Algorithm 15 Invertibility Criterion for relative coordinates

Require: support relset in relative coordinates

Ensure: True only if the Invertibility Criterion is successful; False if inconclusive

```
    1: function invertibility_criterion(relset)
    2: L ← divide(relset)
    3: return all_is_true([conquer(s) for s in make_subrelset(relset, L)])
```

### Algorithm 16 Make subrelset

4: end function

```
1: function MAKE_SUBRELSET(relset, L)
2: \mathbf{i} \leftarrow 0
3: \mathbf{for} \ n \in \{l \in L \mid l > 0\} \ \mathbf{do}
4: \mathbf{yield} \ \mathbf{relset}[\mathbf{i} : \mathbf{i} + n]
5: \mathbf{i} \leftarrow \mathbf{i} + n
6: \mathbf{end} \ \mathbf{for}
7: \mathbf{end} \ \mathbf{function}
```

#### Algorithm 17 Find Relative Coordinates

```
Require: \mathbf{s}' \in H^{\Xi}
Ensure: R(\mathbf{s}')
 1: function Relsets(s')
       constraints ← [make_rel_constraints(t) for t in s' if t != 0]
       accu \leftarrow list()
 3:
       res \leftarrow list()
 4:
       function DFS(i)
 5:
 6:
           if i \ge |constraints| then
              res.append(accu.copy())
 7:
              return
 8:
           end if
 9:
           for x \in constraints[i] do
10:
              accu.append(x)
11:
              dfs(i+1)
12:
              accu.pop()
13:
           end for
14:
       end function
15:
       dfs(0)
16:
17:
       return res
18: end function
```

Proof of correctness. Let  $\mathbf{s}' = (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{f})$ . By Proposition 9.15,  $\mathbf{m} \in R(\mathbf{s}')$  if and only if  $\mathbf{m} \cap C \neq \emptyset$  for all  $C \in \{\text{make\_rel\_constraints}(\mathbf{t}) \mid t \in \{x_{i,j}, y_{i,j}, z_{i,j}, b_i, c_i, f_i\}, t \neq 0\}$ .

It just remains to show that  $\mathbf{m} \in \mathbf{res}$  if and only if  $\mathbf{m}$  satisfies all of these constraints. First note that  $x \in \mathbf{m}$  if and only if  $x \in \text{constraints}[i]$  for some i by line eleven. Since  $\mathbf{m} \in \mathbf{res}$  if and only if  $\mathbf{m}$  satisfies all constraints by line six, the algorithm is correct.  $\square$ 

## Algorithm 18 Check Configurations Against Invertibility Criterion

```
Require: \Lambda
```

Ensure: left\_to\_check: supports not satisfying the Invertibility Criterion

1: left\_to\_check  $\leftarrow$  list()

2: for each  $\mathbf{s}' \in \Lambda$  do

3: res  $\leftarrow$  [invertibility\_criterion(R) for  $R \in$  relsets( $\mathbf{s}'$ )]

4: if  $\neg$ all(res) then

5: left\_to\_check.append( $\mathbf{s}'$ )

```
6: end if
7: end for
8: return left_to_check
```

## Symmetry

It remains to consider 1107 cases thanks to the Invertibility Criterion for relative coordinates. We can use symmetry to reduce the number of cases further. Let  $\mathbf{s}' \in \Lambda$  be one of the 1107 cases. We can apply the following symmetries to  $\mathbf{s}'$  similar to the symmetries in Section 4.4:

$$(12) \cdot (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{f}) := ((x_{j,i})_{i,j=0}^3, (y_{j,i})_{i,j=0}^3, (z_{j,i})_{i,j=0}^3, \mathbf{c}, \mathbf{b}, \mathbf{f}),$$

$$(13) \cdot (\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{f}) := ((z_{3-i,j})_{i,j=0}^3, (y_{3-j,3-i})_{i,j=0}^3, (x_{3-i,j})_{i,j=0}^3, \mathbf{f}, \mathbf{c}, \mathbf{b}).$$

Define the group  $S_3$  generated by these two symmetries. For all weakly valid outcomes  $\mathbf{w} \in \mathbb{Z}^{V_d}$ , the actions  $\sigma \in S_3$  satisfy

$$\sigma \cdot \operatorname{contr}_d'(\operatorname{sign}(\mathbf{w})) = \operatorname{contr}_d'(\operatorname{sign}(\sigma \cdot \mathbf{w})).$$

We can use this symmetry to reduce the number of cases to consider.

**Proposition 9.26.** Let  $d \in \mathbb{N}$  and  $\mathbf{s}' \in H^{\Xi'}$ . If there are no weakly valid outcomes  $\mathbf{w} \in \mathbb{Z}^{V_d}$  of degree d with  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = \sigma \mathbf{s}'$  for some  $\sigma \in S_3 \setminus \{(1)\}$ , then there are no weakly valid outcomes  $\mathbf{v} \in \mathbb{Z}^{V_d}$  of degree d with  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{v})) = \mathbf{s}'$ .

Proof by Contraposition. Let  $\sigma \in S_3 \setminus \{(1)\}$ . Assume there is some valid outcome  $\mathbf{v} \in \mathbb{Z}^{V_d}$  of degree d such that  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{v})) = \mathbf{s}'$ . Define  $\mathbf{w} \coloneqq \sigma \mathbf{v}$ . Then,  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = \sigma \cdot \operatorname{contr}'_d(\operatorname{sign}(\mathbf{v})) = \sigma \mathbf{s}'$ . Note that  $\mathbf{w}$  is weakly valid since the symmetry group  $S_3$  preserves the weak validity of outcomes. This shows the claim.

We take the 1107 cases of  $\mathbf{s}' \in \Lambda$ , apply the symmetry (12) to each case and check if the Invertibility Criterion for relative coordinates is satisfied. The inconclusive cases are then checked with the symmetry (13). We find that 758 cases are impossible and 349 cases are inconclusive. A pseudo code implementation of this symmetry reduction is given in Algorithm 19. More details can be found in Appendix TODO.

Of the 349 configurations, we consider the equivalence relation  $\sim$  defined by  $w \sim v \iff w = (12)v$  or w = (13)v. We compute the equivalence classes of  $\sim$ , and see there are 179 equivalence classes. By Proposition 4.14 and 4.16, it suffices to consider one contracted configuration  $\mathbf{s}'$  from each of the 179 equivalence classes.

## **Hexagon Criterion**

We have made significant progress in reducing the number of cases to consider. We are left with 179 cases of  $\mathbf{s}' \in \Lambda$  out of 2290 initial cases. It is now time to use our third and last tool from our toolbox, the Hexagon Criterion.

Let  $d \geq 42$ . We apply the Hexagon Criterion (Proposition 8.2) with

$$\ell_1 = \ell_2 = 7$$
 and  $d' = 6$ .

First, we check that the requirements of the Hexagon Criterion are met:  $d' + \ell_1 + \ell_2 = 20 \le 42 = d$ . Next, we need to find all contracted configurations from the 179 cases of  $\mathbf{s}' \in \Lambda$  whose support lies outside the hexagon spanned by  $d', \ell_1$  and  $\ell_2$ , see Figure 8.1.

### Algorithm 19 Further Reduciton by Symmetry

```
Require: 1107 inconclusive cases \mathbf{s}' \in \Lambda
Ensure: left_to_check: supports not satisfying the Invertibility Criterion
 1: tmp \leftarrow list(), left_to_check \leftarrow list()
 2: for each \mathbf{s}' \in \{(12)\mathbf{t}' : \mathbf{t}' \in \Lambda\} do
         res \leftarrow [invertibility\_criterion(R) \text{ for } R \in relsets(s')]
         if \negall(res) then
 4:
 5:
             tmp.append((21)s')
 6:
         end if
 7: end for
 8: for each \mathbf{s}' \in \{(13)\mathbf{t}' : \mathbf{t}' \in \mathsf{tmp}\}\ \mathbf{do}
         res \leftarrow [invertibility\_criterion(R) \text{ for } R \in relsets(s')]
 9:
         if \negall(res) then
10:
11:
             left_to_check.append((31)s')
12:
         end if
13: end for
14: return left_to_check
```

**Proposition 9.27.** Let  $d \geq 42$ . Let  $\mathbf{s}' \in H^{\Xi'}$  be some contracted configuration and  $\mathbf{w} \in \mathbb{Z}^{V_d}$  be some configuration with  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = \mathbf{s}'$ . If  $\operatorname{supp}^+(\mathbf{s}') \cap \{\mathbf{b}, \mathbf{c}, \mathbf{f}\} = \emptyset$ , then  $\operatorname{supp}(\mathbf{w})$  lies outside the hexagon spanned by d',  $\ell_1$  and  $\ell_2$ .

*Proof.* This follows immediately from the definition of the hexagon spanned by d',  $\ell_1$  and  $\ell_2$ , and the definition of the map  $\operatorname{contr}'_d$ .

A simple computation shows that 166 cases out of the 179 cases satisfy the Proposition above (for details see Appendix TODO). By the Hexagon Criterion, valid outcomes  $\mathbf{w}$  with  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = \mathbf{s}'$  where  $\mathbf{s}'$  is one of these 166 cases have degree at most d' = 20. However, we already showed that there are no valid outcomes of degree  $\leq 41$ ; we found a contradiction. Thus, it remains to check the thirteen cases where the Hexagon Criterion could not be applied.

Here are the thirteen cases of  $s' \in \Lambda$  that remain:

- (1)  $\{y_{0.3}, z_{2.0}, z_{2.2}, z_{3.1}, c_1\}$
- (2)  $\{y_{0,3}, y_{1,2}, y_{2,1}, z_{3,0}, b_1\}$
- (3)  $\{y_{0,3}, y_{1,1}, y_{1,3}, z_{2,0}, b_1\}$
- $(4) \{y_{0,3}, y_{1,3}, y_{2,2}, z_{2,0}, b_1\}$
- (5)  $\{x_{0,1}, x_{2,1}, y_{1,3}, z_{3,0}, d_1\}$
- (6)  $\{y_{0,3}, z_{1,0}, z_{2,2}, z_{3,1}, c_1\}$

```
(7) \{y_{0,2}, y_{1,1}, y_{1,3}, z_{3,0}, b_1\}
```

(8) 
$$\{y_{0,3}, z_{2,2}, z_{3,0}, z_{3,1}, c_1\}$$

(9) 
$$\{y_{0.3}, z_{1.1}, z_{3.0}, z_{3.1}, c_1\}$$

(10) 
$$\{x_{0,3}, x_{1,1}, x_{3,0}, z_{3,3}, d_0\}$$

(11) 
$$\{x_{0,2}, x_{2,1}, y_{1,3}, z_{3,0}, d_1\}$$

(12) 
$$\{y_{0,3}, z_{3,0}, b_1, c_1, d_1\}$$

(13) 
$$\{x_{1,2}, x_{2,1}, y_{0,3}, z_{3,0}, d_1\}$$

Let  $d \geq 42$ . We can go through these cases one by one and apply the Hexagon Criterion.

Example 9.28. Consider the first case.

```
y03 z20 z22 z31 c1
```

```
у
У
У
        У
У
            У
        У
                d
С
                d
                     d
С
                d
                     d
                         d
                d
                         d
С
                             d
                     d
                         d
                             d
                                 d
С
        С
            С
                         d
                             d
                                  d
                                      d
С
            С
                             d
                                 d
                                      d
                                          d
С
                             b
                                  z
X
                b
                     b
                         b
                         b
                             b
                                  b
Х
        Х
            Х
                b
                     b
                                      z
                                          Z
                                  b
                         b
                             b
                                      b
                b
                     b
                                  b
                                              z
                                                           Z
```

To apply the Hexagon Criterion to some configuration  $\mathbf{w}$  with the support indicated as above, we need to know where the nonzero entry of  $\mathbf{w}$  in the  $c_1$ -column is roughly located; here the  $c_1$ -column denotes the entries  $w_{1,k}$  with  $k=4,\ldots,d-1-4$ . There can only exist one nonzero entry in the  $c_1$ -column since  $\mathbf{w}$  has positive support size five. Let us denote the nonzero entry in the  $c_1$ -column by  $(a,b) \in V_d$  where a=1. We differentiate between three cases:

(1) Let  $a+b \leq \operatorname{floor}(\frac{d}{3})$ . Set  $\ell_1 = \ell_2 = d' = \operatorname{floor}(\frac{d}{3}) \geq \frac{42}{3} = 14$ . We easily see that the other non-entries  $y_{0,3}, z_{2,0}, z_{2,2}, z_{3,1}$  lie outside the hexagon.

(2) Let  $b \ge \text{floor}(\frac{d}{3})$ . Set d' = 6,  $\ell_2 = 7$  and  $\ell_1 = d + 1 - \text{floor}(\frac{d}{3}) \ge 43 - 14 = 29$ . We see that  $d' + \ell_1 + \ell_2 \le d$  since  $d \ge 42$ . We easily see that the non-entries  $y_{0,3}, z_{2,0}, z_{2,2}, z_{3,1}$  lie outside the hexagon. It remains to check that (a, b) also lies outside the hexagon. We have  $d - \ell_1 = d - (d + 1) + \text{floor}(\frac{d}{3}) = \text{floor}(\frac{d}{3}) - 1 < \text{floor}(\frac{d}{3}) \le b$ . Thus, (a, b) lies outside the hexagon.

Both these cases show that we can apply the Hexagon Criterion; the criterion shows that the degree of **w** is at most  $d' \leq \text{floor}(\frac{d}{3})$ . However, we assume that **w** is of degree d, which is a contradiction since d' < d. We rule out  $\{y_{0,3}, z_{2,0}, z_{2,2}, z_{3,1}, c_1\}$ .

We generalize Example 9.28 to apply the Hexagon Criterion to the remaining twelve cases. Unfortunately, there is one case  $\{y_{0,3}, z_{3,0}, b_1, c_1, d_1\}$  which cannot be ruled out by the Hexagon Criterion. We need to consider this case separately; so let us ignore this case for now and consider the remaining eleven cases.

First, we observe that all the eleven cases satisfy

$$|\sup^+(\mathbf{s}') \cap \{b_0, b_1, c_0, c_1, f_0, f_1\}| = 1$$
 and  $|\sup^+(\mathbf{s}') \cap \{b_2, c_2, f_2, b_3, c_3, f_3\}| = 0$ .

Hence, for all valid outcomes  $\mathbf{w} \in \mathbb{Z}^{V_d}$  with  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = \mathbf{s}'$  there exists some vertex  $(a,b) \in V_d$  with a=0,1 or b=0,1 or  $a+b \geq d-1$  such that

$$supp(\mathbf{w}) \setminus \{(a,b)\} \subset V_6 \cup \{(i,j) \in V_d \mid j > d-7\} \cup \{(i,j) \in V_d \mid i > d-7\}.$$

**Proposition 9.29.** One of the following statements holds:

- (1)  $a + b \leq \text{floor}(\frac{d}{3}),$
- (2)  $a \ge \text{floor}(\frac{d}{3}), or$
- (3)  $b \ge \text{floor}(\frac{d}{3})$ .

*Proof.* We make a case distinction:

- Let  $a + b \le \text{floor}(\frac{d}{3})$ . Our claim holds.
- Let a = 0 with  $a + b > \text{floor}(\frac{d}{3})$ . Then,  $b > \text{floor}(\frac{d}{3})$ . Let a = 1 with  $a + b > \text{floor}(\frac{d}{3})$ . Then,  $b \ge \text{floor}(\frac{d}{3})$ .
- Let b = 0 with  $a + b > \text{floor}(\frac{d}{3})$ . Then,  $a > \text{floor}(\frac{d}{3})$ . Let b = 1 with  $a + b > \text{floor}(\frac{d}{3})$ . Then,  $a \ge \text{floor}(\frac{d}{3})$ .
- Let  $a+b \ge d-1$ . Since  $d-1 > 2\text{floor}(\frac{d}{3})$ , we have  $a > \text{floor}(\frac{d}{3})$  or  $b > \text{floor}(\frac{d}{3})$ .

This proves the proposition.

**Proposition 9.30.** All the eleven cases are impossible.

*Proof.* It is easy to see that all of the eleven cases either satisfy 1., 2., or 3. of Proposition 9.29. When 1. holds, we let  $d' = \ell_1 = \ell_2 = \text{floor}(\frac{d}{3}) \ge 14 > 6$ . When 2. holds, we choose  $d' = 6, \ell_1 = 7$  and  $\ell_2 = d + 1 - \text{floor}(\frac{d}{3})$ . Note that  $d' + \ell_1 + \ell_2 \le 42$  since  $d \ge 42$ . When 3. holds, we choose  $d' = 6, \ell_2 = 7$  and  $\ell_1 = d + 1 - \text{floor}(\frac{d}{3})$ .

We apply the Hexagon Criterion with these choices of d',  $\ell_1$  and  $\ell_2$  to each of the eleven cases. We find that the degree of the valid outcomes  $\mathbf{w}$  is at most d'. However, we assumed that the degree of the valid outcomes is d, which is a contradiction. This shows that all of the eleven cases are impossible.

#### The Final Case

We are left with one case  $\{y_{0,3}, z_{3,0}, b_1, c_1, d_1\}$ .

**Proposition 9.31.** There is no weakly valid outcome  $\mathbf{w} \in \mathbb{Z}^{V_d}$  such that  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = \{x_{0,0}, y_{0,3}, z_{3,0}, b_1, c_1, d_1\}.$ 

Proof by Contradiction. We assume that  $d \geq 42$ . Assume that a weakly valid outcome  $\mathbf{w} \in \mathbb{Z}^{V_d}$  with  $\operatorname{contr}'_d(\operatorname{sign}(\mathbf{w})) = \{x_{0,0}, y_{0,3}, z_{3,0}, b_1, c_1, d_1\}$  exists. Then, the support of  $\mathbf{w}$  reads

$$S = \{(0,0), (d,0), (0,d), (i,1), (1,j), (k,d-1-k)\}$$

for some integers i, j, k. Let us define  $e \in \mathbb{Z}$  such that we can write d = 2e + 1. If  $j \neq e$ , we use Proposition 4.20 to show that such a **w** cannot exist. So, we assume that j = e. By symmetry  $(12) \in S_3$  and  $(13) \in S_3$ , we conclude that i = e and k = e. This shows that the support of **w** reads

$$S = \{(0,0), (d,0), (0,d), (e,1), (1,e), (e,d-1-e)\}$$
  
= \{(0,0), (d,0), (0,d), (e,1), (1,e), (e,e)\}.

We apply the Invertibility Criterion with  $E := \{0, 1, 3, e, d - 1, d\}$  which leads to the following pairing matrix

$$A_{E,S}^{(d)} = \begin{bmatrix} 1 & 0 & 1 & 0 & 0 & 0 \\ d & 0 & 0 & 0 & 1 & 0 \\ \binom{d}{3} & 0 & 0 & 0 & \binom{e}{2} & 0 \\ \binom{e}{d} & 0 & 0 & 1 & e & 1 \\ d & 0 & 0 & 1 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 \end{bmatrix}.$$

This pairing matrix has determinant  $\frac{(2e+1)(e+1)e}{6} \neq 0$ . Thus, S cannot be the support of a valid nonzero outcome.

With the last case excluded, we proved that for all valid integral outcomes  $\mathbf{w}$  with  $|\sup_{\mathbf{v}}^{+}(\mathbf{w})| = 5$  we have

$$\deg(\mathbf{w}) \le 7.$$

# Chapter 10

# Valid Outcomes of Positive Support Size Six

We want to prove that for all valid integral outcomes  $\mathbf{w}$  with  $|\operatorname{supp}^+(\mathbf{w})| = 6$  we have

$$\deg(\mathbf{w}) \leq 9.$$

The original paper by Bik and Marigliano [2] did not attempt to prove this result. This thesis provides a significant step towards proving this result by reducing the number of cases to check, which renders the problem computationally feasible.

In principle, we could use the same idea as in the previous chapters to find all valid outcomes of positive support size six; we start with computing  $\Gamma^{\text{even}} \cap \{ \mathbf{s} \in H^{\Xi} : |\text{supp}^+(\mathbf{s})| = 6 \}$  and  $\Gamma^{\text{odd}} \cap \{ \mathbf{s} \in H^{\Xi} : |\text{supp}^+(\mathbf{s})| = 6 \}$  albeit with a slight modification to these sets. Whereas we have previously considered contractions of size four, we now consider contractions of size five, i.e. we have the contraction variables as depicted in Figure 10.1. From now on, when we write  $\mathbf{s} \in H^{\Xi}$  we mean contractions of size five.

#### **Definition 10.1.** We define

$$\Gamma_6^{\text{even}} := \Gamma^{\text{even}} \cap \left\{ \mathbf{s} \in H^{\Xi} : |\text{supp}^+(\mathbf{s})| = 6 \right\} \text{ and } \Gamma_6^{\text{odd}} := \Gamma^{\text{odd}} \cap \left\{ \mathbf{s} \in H^{\Xi} : |\text{supp}^+(\mathbf{s})| = 6 \right\},$$

where  $\Gamma^{\text{even}}$  and  $\Gamma^{\text{odd}}$  are defined analogously to Definition 7.1, but now they contain hyperfield configurations of contraction size five for

$$\Phi_1 := \{ \text{col}(1), \text{col}(2), \text{col}(3), \text{col}(3), \text{row}(1), \text{row}(2), \text{row}(3), \text{row}(4), \\ \text{diag}(1), \text{diag}(2), \text{diag}(3), \text{diag}(4), \text{diag}(d-1), \text{diag}(d-2), \text{diag}(d-3), \text{diag}(d-4) \},$$

and

$$\Phi_2 := \{ \operatorname{col}(d), \operatorname{col}(d-1), \operatorname{col}(d-2), \operatorname{col}(d-3), \operatorname{col}(d-4), \operatorname{row}(d), \operatorname{row}(d-1), \operatorname{row}(d-2), \operatorname{row}(d-3), \operatorname{row}(d-4) \}.$$

```
y_{0,4}
y_{0,3}
          y_{1,4}
y_{0,2}
          y_{1,3}
                    y_{2,4}
y_{0,1}
          y_{1,2}
                     y_{2,3}
                               y_{3,4}
y_{0,0}
          y_{1,1}
                    y_{2,2}
                               y_{3,3}
                                         y_{4,4}
                                                   d_0
 c_0
                                         y_{4,3}
          y_{1,0}
                     y_{2,1}
                               y_{3,2}
                                                   d_1
 c_0
           c_1
                     y_{2,0}
                               y_{3,1}
                                         y_{4,2}
                                                           e_0
                                                   d_2
                                                           e_1
                                                                   d_0
                                         y_{4,1}
            c_1
                      c_2
                               y_{3,0}
 c_0
                                                   d_3
                                                                   d_1
                                          y_{4,0}
                                                           e_2
                                                                          e_0
 c_0
           c_1
                      c_2
                                c_3
                                                    d_4
 c_0
            c_1
                      c_2
                                c_3
                                           c_4
                                                           e_3
                                                                   d_2
                                                                          e_1
                                                                                   d_0
                                                                   d_3
                                                           e_4
                                                                          e_2
                                                                                   d_1
 c_0
            c_1
                      c_2
                                c_3
                                           c_4
                                                                   d_4
                                           c_4
                                                                          e_3
                                                                                   d_2
                                                                                             e_1
                                                                                                       d_0
 c_0
            c_1
                      c_2
                                c_3
                                                                    *
                                                                           e_4
                                                                                   d_3
                                                                                                       d_1
                                           c_4
            c_1
                      c_2
                                c_3
                                                                                             e_2
                                                                                                                  e_0
 c_0
                                                    b_4
                                                                   b_4
                                         x_{4.4}
                                                                                  z_{0.4}
                                                                                            z_{1.4}
x_{0.4}
          x_{1,4}
                    x_{2.4}
                               x_{3.4}
                                                                                                      z_{2.4}
                                                                                                                z_{3.4}
                                                                                                                          z_{4.4}
                                                   b_3
                                                           b_3
                                                                   b_3
                                                                           b_3
                                                                                   b_3
x_{0,3}
          x_{1,3}
                    x_{2,3}
                               x_{3,3}
                                         x_{4,3}
                                                                                             z_{0,3}
                                                                                                      z_{1,3}
                                                                                                                z_{2,3}
                                                                                                                          z_{3,3}
                                                                                                                                    z_{4,3}
                                                           b_2
                                                    b_2
                    x_{2,2}
                               x_{3,2}
                                         x_{4,2}
                                                                                   b_2
                                                                                                      z_{0,2}
                                                                                                                z_{1,2}
                                                                                                                                    z_{3,2}
                                                                                                                                              z_{4,2}
x_{0,2}
         x_{1,2}
                                                                                                                          z_{2,2}
                                                                                   b_1
                                                    b_1
                                                           b_1
                                                                   b_1
                                                                                                       b_1
                                                                                                                z_{0,1}
x_{0,1}
          x_{1,1}
                    x_{2,1}
                               x_{3,1}
                                         x_{4,1}
                                                                                                                          z_{1,1}
                                                                                                                                    z_{2,1}
                                                                                                                                              z_{3,1}
                                                                                                                                                        z_{4,1}
                                                    b_0
                                                                                                        b_0
x_{0,0}
                                                           b_0
                                                                                                                  b_0
                                                                                                                          z_{0,0}
          x_{1.0}
                    x_{2.0}
                               x_{3.0}
                                         x_{4.0}
                                                                                                                                    z_{1,0}
                                                                                                                                              z_{2,0}
                                                                                                                                                        z_{3,0}
                                                                                                                                                                  z_{4,0}
```

Figure 10.1: Contraction variables for d = 16 are depicted.

We increased the contraction size to gain more information about potential supports, which we hope will prove useful when handling cases manually.

#### **Proposition 10.2.** The following holds:

```
(1) We have |\Gamma_6^{\text{even}}| = 150032.
```

(2) We have  $|\Gamma_6^{\text{odd}}| = 154177$ .

*Proof.* This is verifed by computer. TODO.

The number of cases to check has increased by a factor of hundred compared to the previous chapters. If we would be able to reduce the cases immensely, we could in theory apply the same techniques as in the previous chapters to prove  $\deg(\mathbf{w}) \leq 9$ . We will spend the remaining part of this chapter to find a way to reduce the number of cases to check to around 10,000 cases; this is a reduction by a factor of 30. We hope that this reduction is sufficient to make the problem computationally feasible for future proofs. Due to time constraints of this thesis, we will not be able to prove  $\deg(\mathbf{w}) \leq 9$  for all valid outcomes of positive support size six.

## 10.1 Fixed-Contractable Pascal Forms

To reduce the number of cases to check, we want to generate a set of hyperfield Pascal forms that are *fixed-contractable*.

**Definition 10.3.** Let  $d \in \mathbb{N}_{\geq 15}$  and p be a hyperfield linear form on  $H^{V_d}$ . We say p is contractable for d if we can write  $p = \hat{p}$  for some linear form  $\hat{p} \in H[\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}]$ .

**Definition 10.4.** Let  $d \in \mathbb{N}_{\geq 15}$ ,  $k = 0, \ldots, 4$ , and p be a hyperfield linear form on  $H^{V_d}$ . We say p is *contractable* for d on  $b_k$  if we can write

$$p = \sum_{(i,j) \in V_d \setminus \{(5,k),\dots,(d-k-5,k)\}} \lambda_{i,j} x_{i,j} + \lambda b_k \quad \text{for} \quad b_k := x_{5,k} + \dots + x_{d-5-k,k}$$

for  $\lambda_{i,j}$ ,  $\lambda \in H$ . In other words, we can write  $p = \hat{p}$  for some linear form  $\hat{p} \in H[\mathbf{x}, b_k]$ . Similarly, we define p is contractable on  $c_k$ ,  $d_k$ , and  $e_k$  if we can write  $p = \hat{p}$  for some linear form  $\hat{p} \in H[\mathbf{x}, c_k]$ ,  $\hat{p} \in H[\mathbf{x}, d_k]$ , and  $\hat{p} \in H[\mathbf{x}, e_k]$ , respectively.

**Remark 10.5.** Clearly, p is contractable for d if and only if it is contractable for d on  $b_k$ ,  $c_k$ ,  $d_k$ , and  $e_k$  for all  $k = 0, \ldots, 4$ .

We are interested in linear combinations of hyperfield Pascal forms p that are contractable such that  $\hat{p}$  is independent of the degree d. We call such linear combinations fixed-contractable because  $\hat{p}$  is fixed over all degrees d. To formalize this, we introduce the following definition.

**Definition 10.6.** Let t be a formal variable and T be a formal linear combination of

$$\left\{ {\rm sign}({\rm diag}(k)), {\rm sign}({\rm row}(k)), {\rm sign}({\rm col}(k)) \right\}_{k \in \{0,1,2,3,4,t-4,t-3,t-2,t-1,t\}}.$$

Let  $d \in \mathbb{N}_{\geq 15}$ . We write  $p_d$  for the realization of T for t = d; the realization  $p_d$  is just a linear form on  $H^{V_d}$  where we replace the formal variable t by an actual value d.

**Example 10.7.** Consider the formal linear combination T = row(3) - row(t-2). Let d = 15. The realization  $p_d$  of T for t = d is the linear form depicted in the following figure

```
-350
-350
-285
        65
              65
-220
        65
                    -65
-165
        55
             -10
                   -10
                           55
-120
             -10
        45
                           10
                                -45
 -84
        36
               -9
                      1
                                 -9
                                        36
 -56
        28
                      1
                                 -1
                                         8
               -8
                                             -28
 -35
        21
               -7
                                         1
                                              -7
                                                    21
 -20
        15
               -6
                      1
                                              -1
                                                     6
                                                         -15
 -10
        10
               -5
                      1
                                                                 10
  -4
         6
               -4
                                                                       -6
         3
                                                                       -3
  -1
               -3
                                                                               3
         1
                                                                               2
               -2
                      1
                                                                                    -1
                                                                                    -1
                                                                               1
               -1
                                                                                    -1
```

The realization  $p_d$  is contractable for all  $d \in \mathbb{N}$  with  $d \geq 15$ , but does the linear form  $\hat{p}_d \in H[\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}]$  change with d? We will show that it does not change.

**Definition 10.8.** Let T be a formal linear combination of

```
\left\{ {\rm sign}({\rm diag}(k)), {\rm sign}({\rm row}(k)), {\rm sign}({\rm col}(k)) \right\}_{k \in \{0,1,2,3,4,t-4,t-3,t-2,t-1,t\}}.
```

We say T is fixed-contractable if all of the following statements hold:

- (1) the realization  $p_d$  is contractable for all  $d \in \mathbb{N}$  with  $d \in \mathbb{N}_{\geq 15}$ ;
- (2) there exists a linear form  $\hat{p}^{\text{even}} \in H[\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}]$  such that  $p_d = \hat{p}^{\text{even}}$  for all even degrees  $d \in \mathbb{N}_{\geq 15}$ ;
- (3) there exists a linear form  $\hat{p}^{\text{odd}} \in H[\mathbf{x}, \mathbf{y}, \mathbf{z}, \mathbf{b}, \mathbf{c}, \mathbf{d}, \mathbf{e}]$  such that  $p_d = \hat{p}^{\text{odd}}$  for all odd degrees  $d \in \mathbb{N}_{>15}$ .

Similarly, we define p is fixed-contractable on  $b_i$  if  $\hat{p}^{\text{even}} \in H[\mathbf{x}, b_i]$  such that  $p_d = \hat{p}^{\text{even}}$  for all even degrees  $d \in \mathbb{N}_{\geq 15}$  and  $\hat{p}^{\text{odd}} \in H[\mathbf{x}, b_i]$  such that  $p_d = \hat{p}^{\text{odd}}$  for all odd degrees  $d \in \mathbb{N}_{\geq 15}$ .

The definition of fixed-contractable on  $c_i$ ,  $d_i$ , and  $e_i$  is analogous.

**Example 10.9.** By Proposition 6.31 and 6.33, the following formal linear combinations are fixed-contractable: col(1), col(2), col(3), row(1), row(2), row(3), diag(1), diag(2), diag(3),

diag(d-1), diag(d-2), diag(d-3), col(d), col(d-1), col(d-2), col(d-3), row(d), row(d-1), row(d-2), row(d-3).

We are looking for more fixed-contractable formal linear forms.

**Definition 10.10.** Let  $d \in \mathbb{N}_{\geq 15}$ ,  $i \in \{0, \dots, 4, d-4, \dots, d\}$ , and  $p = \sum \lambda_{i,j} x_{i,j}$  be a hyperfield linear form on  $H^{V_d}$ . We define the *i*-th *b*-row of p as  $p_{b_i} := \begin{bmatrix} \lambda_{5,i} & \dots & \lambda_{d-i-5,i} \end{bmatrix} \in H^{d-i-9}$ . Similarly, we define  $p_{c_i}$ ,  $p_{d_i}$  and  $p_{e_i}$  to denote the *i*-th *b*-column, *d*-diagonal and *e*-diagonal of p, respectively.

**Proposition 10.11.** Let  $d \in \mathbb{N}_{\geq 15}$ ,  $i = 0, \ldots, 4$ , and T be a formal linear combination of  $\{\operatorname{sign}(\operatorname{row}(j)), \operatorname{sign}(\operatorname{col}(j)), \operatorname{sign}(\operatorname{diag}(j)) \mid j \in \{0, \ldots, 4\} \cup \{t-4, \ldots, t\}\}$ . Then, the following statements hold for all realizations  $p_d$  of T:

- (1) The c-column of  $(p_d)_{c_i}$  only depends on  $\{\operatorname{sign}(\operatorname{row}(k)), \operatorname{sign}(\operatorname{diag}(k))\}_{k=0}^4$ .
- (2) The b-row of  $(p_d)_{b_i}$  only depends on  $\{\operatorname{sign}(\operatorname{col}(k)), \operatorname{sign}(\operatorname{diag}(d-k))\}_{k=0}^4$ .
- (3) The d-diagonal of  $(p_d)_{d_i}$  only depends on  $\{\operatorname{sign}(\operatorname{row}(d-k)), \operatorname{sign}(\operatorname{col}(d-k))\}_{k=0}^4$ . A similar statement holds for e-diagonals.

*Proof.* This follows immediately from the definition of row, col, and diag.  $\Box$ 

**Proposition 10.12.** Let  $d \in \mathbb{N}_{\geq 15}$  and i = 0, ... 4. The following statements hold:

- (1) Let  $T \in \{\operatorname{sign}(\operatorname{row}(k)), \operatorname{sign}(\operatorname{diag}(k))\}_{k=0,\dots,4}$ . The  $c_i$ -column of  $p_d$  is a constant vector.
- (2) Let  $T \in \{\operatorname{sign}(\operatorname{col}(k)), \operatorname{sign}(\operatorname{diag}(d-k))\}_{k=0,\dots,4}$ . The  $b_i$ -row of  $p_d$  is a constant vector.
- (3) Let  $T \in \{ sign(row(d-k)), sign(col(d-k)) \}_{k=0,\dots,4}$ . The  $d_i$ -diagonal of  $p_d$  is a constant vector; similarly for the  $e_i$ -diagonal.

*Proof.* This also follows easily from the definition of row, col, and diag.  $\Box$ 

Let us investigate how realizations of certain formal linear combinations change when we increase the degree d.

**Example 10.13.** Consider T = row(3) + row(2). Then,  $p_8$  is represented by the triangle on the left and  $p_9$  is represented by the triangle on the right.

							-48						
-28							-28	20					
-14	14						-14	14	-6				
-5	9	-5					-5	9	-5	1			
	5	-4	1				•	5	-4	1			
2	2	-3	1				2	2	-3	1			
2		-2	1				2		-2	1			
1	-1	-1	1				1	-1	-1	1			
	-1		1					-1		1			
		1	1		•				1	1		•	

For the following propositions and lemmas, we fix the following notations:

• Let T be a formal linear combination of  $\{\text{row}(i)\}_{i=0}^4$  or  $\{\text{row}(i)\}_{i=t-4}^t$ , i.e.

$$T = \sum_{i=0}^{4} \lambda_i \operatorname{row}(i)$$
 or  $T = \sum_{i=0}^{4} \lambda_{t-i} \operatorname{row}(t-i);$ 

- Write the realizations of T as  $p := \sum p_{i,j} x_{i,j} := p_d$  and  $q := \sum q_{i,j} x_{i,j} := p_{d+1}$  for some degree t = d;
- Define  $r := \max\{i \mid \lambda_i \neq 0\}$  if  $\max\{i \mid \lambda_i \neq 0\} \leq 4$ , otherwise  $r := \min\{i \mid \lambda_i \neq 0\}$ ;
- Write  $row(r) = \sum r_{i,j} x_{i,j}$ .

**Lemma 10.14.** Let  $d \in \mathbb{N}_{\geq 9}$ . Then,  $q_{i,j} = p_{i,j}$  holds for all  $(i,j) \in V_d$ .

*Proof.* Let  $d \in \mathbb{N}$ ,  $\ell = 0, \ldots, 4, t - 4, \ldots, t$  and  $T_{\ell} = \text{row}(\ell)$ . We denote its realizations in d and d+1 by  $p_d^{(\ell)} = \sum p_{i,j}^{(\ell)} x_{i,j}$  and  $p_{d+1}^{(\ell)} = \sum q_{i,j}^{(\ell)} x_{i,j}$ , respectively. By Proposition 3.18, we see that  $p_{i,j}^{(\ell)} = q_{i,j}^{(\ell)}$  for all  $(i,j) \in V_d$ .

Next, assume 
$$T = \sum_{i=0}^4 \lambda_i \operatorname{row}(i) + \sum_{i=0}^4 \lambda_{t-i} \operatorname{row}(t-i)$$
. Then,  $p_{i,j} = \sum_{\ell} \lambda_{\ell} p_{i,j}^{(\ell)} = \sum_{\ell} \lambda_{\ell} q_{i,j}^{(\ell)} = q_{i,j}$  for all  $(i,j) \in V_d$ .

**Lemma 10.15.** Let  $d \in \mathbb{N}_{\geq 9}$ . If there exists  $k \in \{0, \ldots, r\}$  such that  $\operatorname{sign}(r_{i,d-i}) = \operatorname{sign}(p_{i,d-i})$  for all  $i = k, \ldots, r$ , then  $\operatorname{sign}(q_{i,d+1-i}) = \operatorname{sign}(p_{i,d-i})$  for all  $i = k, \ldots, r$ .

*Proof.* Without loss of generality, we assume that  $\lambda_r > 0$ . First, we see that  $q_{r,\cdot} := (q_{r,j})_{j=0}^{d-r} = \lambda_r \cdot \mathbf{1}$  and  $q_{i,\cdot} = \mathbf{0}$  for all i > r. By the Pascal property, we have

$$q_{r-1,d+1-(r-1)} = q_{r-1,d-(r-1)} - q_{r,d+1-r} = q_{r-1,d-(r-1)} - \lambda_r.$$

This shows  $q_{r-1,d+1-(r-1)} < q_{r-1,d-(r-1)} = p_{r-1,d-(r-1)} < 0$ , where the last inequality follows from the assumption  $\operatorname{sign}(r_{i,d-i}) = \operatorname{sign}(p_{i,d-i})$ . Thus, we have  $\operatorname{sign}(q_{r-1,d+1-(r-1)}) = \operatorname{sign}(q_{r-1,d-(r-1)}) = -1$ .

Next, we again use the Pascal property  $q_{r-2,d+1-(r-2)} = q_{r-2,d-(r-2)} - q_{r-1,d+1-(r-1)}$ . We see that  $q_{r-2,d+1-(r-2)} > 0$  because  $q_{r-2,d-(r-2)} > 0$  and  $q_{r-1,d+1-(r-1)} < 0$ . Thus, we have  $sign(q_{r-2,d+1-(r-2)}) = sign(q_{r-2,d-(r-2)}) = 1$ . We can continue this argument for  $r-3, r-4, \ldots, k$ . This shows that  $sign(q_{i,d+1-i}) = sign(q_{i,d-i}) = sign(p_{i,d-i})$  for all  $i = k, \ldots, r$ .

**Lemma 10.16.** Let  $d \in \mathbb{N}_{\geq 15}$  and  $r \leq 4$ . If there exists  $k \in \{0, \ldots, r\}$  such that for all  $i = k, \ldots, r$  we have  $\operatorname{sign}(\operatorname{row}(r))_{c_i} = \operatorname{sign}(p)_{c_i}$ , then for all  $i = k, \ldots, r$  we either have  $p_{c_i} > 0, q_{c_i} > 0$  or  $p_{c_i} < 0, q_{c_i} < 0$ .

*Proof.* Let i = k, ..., r. If we show  $\operatorname{sign}(p_{i,d-i}) = \operatorname{sign}(p_{i,d-i-5})$ , then we have  $\operatorname{sign}(r_{i,d-i}) = \operatorname{sign}(p_{i,d-i})$  by assumption  $\operatorname{sign}(\operatorname{row}(r))_{c_i} = \operatorname{sign}(p)_{c_i}$ . So, we can use Lemma 10.15 to prove the statement.

It is easy to see that  $\operatorname{sign}(p_{r,d-r}) = \operatorname{sign}(p_{r,d-r-5})$  since  $\operatorname{sign}(\operatorname{row}(r))_{c_r} = \operatorname{sign}(p)_{c_r}$  and  $\operatorname{sign}(r_{r,d-r}) = \operatorname{sign}(r_{r,d-r-5})$ . For r-1, we then use the Pascal property. We repeat this argument for  $r-2, r-3, \ldots, k$ . This shows that  $\operatorname{sign}(p_{i,d-i}) = \operatorname{sign}(p_{i,d-i-5})$  for all  $i = k, \ldots, r$ .

**Proposition 10.17.** Let  $d \in \mathbb{N}_{\geq 15}$  and  $r \leq 4$ . If there exists  $k \in \{0, \ldots, r\}$  such that for all  $i = k, \ldots, r$  we have  $\operatorname{sign}(\operatorname{row}(r))_{c_i} = \operatorname{sign}(p)_{c_i}$ , then  $\operatorname{sign}(T)$  is fixed-contractable on  $c_i$  for all  $i = k, \ldots, r$ .

Proof. Let i = k, ..., r and  $d \in \mathbb{N}_{\geq 15}$  First, it is easy to see that p is contractable on  $c_i$  because  $\operatorname{row}(r)$  is contractable and  $\operatorname{sign}(\operatorname{row}(r))_{c_i} = \operatorname{sign}(p)_{c_i}$ . By Lemma 10.16 the sign does not change when increasing the degree  $d \leadsto d+1$ . Hence, the contractability of p on  $c_i$  is preserved for all degrees greater or equal to d. Therefore, there exists one  $\hat{p} \in H[\mathbf{x}, c_i]$  for all  $d' \geq d$  such that  $\hat{p} = p_{d'}$ .

There exist similar propositions for contractability on  $d_i$  and  $e_i$ .

**Proposition 10.18.** Let  $d \in \mathbb{N}_{\geq 15}$  and  $r \geq d-4$ . If there exists  $k \in \{r, \ldots, d\}$  such that for all  $i = r, \ldots, k$  we have  $\operatorname{sign}(\operatorname{row}(r))_{d_i} = \operatorname{sign}(p)_{d_i}$ , then  $\operatorname{sign}(T)$  is fixed-contractable on  $d_i$  for all  $i = r, \ldots, k$ . The analogous statement holds for  $e_i$ .

*Proof.* We can use the same proof as before, but now the sign of the entire diagonal  $d_i$  changes whenever we increase the dimension by one. Fortunately, the contractability on  $d_i$  is not affected by this.

We state analogous propositions for  $\operatorname{col}(\cdot)$  of Proposition 10.17 and 10.18 but skip the proofs since they are similar. Let  $T = \sum_{i=0}^{4} \lambda_i \operatorname{col}(i) + \sum_{i=0}^{4} \lambda_{t-i} \operatorname{col}(t-i)$ .

**Proposition 10.19.** Let  $d \in \mathbb{N}_{\geq 15}$  and  $r \leq 4$ . If there exists  $k \in \{0, \ldots, r\}$  such that for all  $i = k, \ldots, r$  we have  $\operatorname{sign}(\operatorname{col}(r))_{b_i} = \operatorname{sign}(p)_{b_i}$ , then  $\operatorname{sign}(T)$  is fixed-contractable on  $b_i$  for all  $i = k, \ldots, r$ .

**Proposition 10.20.** Let  $d \in \mathbb{N}_{\geq 15}$  and  $r \geq d-4$ . If there exists  $k \in \{r, \ldots, d\}$  such that for all  $i = r, \ldots, k$  we have  $\operatorname{sign}(\operatorname{col}(r))_{d_i} = \operatorname{sign}(p)_{d_i}$ , then  $\operatorname{sign}(T)$  is fixed-contractable on  $d_i$  for all  $i = r, \ldots, k$ . The analogous statement holds for  $e_i$ .

Here is an analogous version of Lemma 10.14 but for diag( $\cdot$ ).

**Lemma 10.21.** Let  $d \in \mathbb{N}_{\geq 9}$ . Then,  $q_{i,j+1} = p_{i,j}$  holds for all  $(i,j) \in V_d$ .

*Proof.* Just use Lemma 10.14 and symmetries  $\sigma \in S_3$ .

**Example 10.22.** Consider T = diag(3) + diag(2). Then,  $p_8$  is represented by the triangle on the left and  $p_9$  is represented by the triangle on the right.

1 1 1 10 6 10 20 10 35 35 15 21 56 84 28 120

Not surprisingly, we have analogous propositions for  $\operatorname{diag}(\cdot)$  of Proposition 10.17. Consider the formal linear combination  $T = \sum_{i=0}^{4} \lambda_i \operatorname{diag}(i) + \sum_{i=0}^{4} \lambda_{t-i} \operatorname{diag}(t-i)$ .

**Proposition 10.23.** *Let*  $d \in \mathbb{N}_{>15}$ . *The following statements hold:* 

- (1) Assume  $r \leq 4$ . If there exists  $k \in \{0, ..., r\}$  such that for all i = k, ..., r we have  $\operatorname{sign}(\operatorname{diag}(r))_{c_i} = \operatorname{sign}(p)_{c_i}$ , then  $\operatorname{sign}(T)$  is fixed-contractable on  $c_i$  for all i = k, ..., r.
- (2) Assume  $r \geq d-4$ . If there exists  $k \in \{r, \ldots, d\}$  such that for all  $i = r, \ldots, d$  we have  $\operatorname{sign}(\operatorname{diag}(r))_{b_i} = \operatorname{sign}(p)_{b_i}$ , then  $\operatorname{sign}(T)$  is fixed-contractable on  $b_i$  for all  $i = r, \ldots, d$ .

*Proof.* The proofs are analogous to the proof of Proposition 10.17.

We define  $G_{\text{row}}$  to be the set of formal linear combinations with

$$G_{\text{row}} := \left\{ \sum_{i=1}^{5} \lambda_{i} \text{row}(k_{i}) \mid \lambda_{i} \in \{-1, 0, 1\}, k_{i} \in \{0, \dots, 4, d-4, \dots, d\} \right\}.$$

The sets  $G_{\text{col}}$  and  $G_{\text{diag}}$  are defined similarly. Using Proposition 10.17 and 10.18 we can easily compute a subset  $G'_{\text{row}} \subset G_{\text{row}}$  that only contains fixed-contractable forms. Refer to GITHUB TODO.

# 10.2 An Extended Trivial System

To compute  $\Gamma_6^{\text{even}}$ , we defined the system  $\Phi = \Phi_1 \cup \Phi_2$ . We proved that this system is non-trivial. If we can find a system  $\Psi$  that is a superset of  $\Phi$  and is also nont-rivial, then we can reduce the number of cases to check.

**Proposition 10.24.** Define  $\Psi' = \Phi \cup \{ \operatorname{diag}(i) - \operatorname{diag}(j) \mid (i, j) \in \mathbb{Z} \}$ , where

$$\begin{split} Z \coloneqq \{(0,1), (0,2), (0,3), (0,4), (0,d-1), (0,d-2), (0,d-3), (0,d-4), \\ (1,2), (1,3), (1,d), (1,d-4), (1,d-2), (1,d-3), \\ (2,d), (2,d-1), (2,d-3), (2,d-4), \\ (3,d), (3,d-1), (3,d-2), (3,d-4) \\ (d-4,d), (d-3,d), (d-2,d), (d-2,d-1), (d-1,d-2), (d-1,d-3), (d-1,d)\}. \end{split}$$

The system  $\Psi'$  is non-trivial.

Proof. It is easy to see that every  $p = \sum \lambda_{i,j} x_{i,j} = \operatorname{diag}(q) - \operatorname{diag}(r)$  for  $(q,r) \in Z$  satisfies  $\lambda_{0,0} \neq 0$ ,  $\operatorname{supp}^+(p) \neq \emptyset$ , and  $\operatorname{supp}^-(p) \neq \emptyset$ . Thus, any root  $\mathbf{w}$  of p satisfies  $\operatorname{supp}^+(p) \cap \operatorname{supp}^+(w) \neq \emptyset$  or  $\operatorname{supp}^-(p) \cap \operatorname{supp}^+(w) \neq \emptyset$ .

**Proposition 10.25.** Let  $T \in \{\operatorname{diag}(i) - \operatorname{diag}(j) \mid (i, j) \in Z\}$ . Then, T is fixed-contractable.

*Proof.* Let  $(i, j) \in \mathbb{Z}$ . Use Proposition 10.23 if  $i, j \leq 4$  or  $i, j \geq d-4$ . Otherwise, the claim follows immediately from Proposition 10.11.

**Proposition 10.26.** Define  $\Psi := \Psi \cup \{ \operatorname{diag}(1) - \operatorname{diag}(d-1) \}$ . The system  $\Psi$  is non-trivial.

*Proof.* Let T = diag(1) - diag(d-1) be a formal linear combination. Let  $d \in \mathbb{N}_{\geq 15}$  be odd and  $\mathbf{w}$  be a root of  $\Psi$ . Then, the realization  $p_d = \sum \lambda_{i,j} x_{i,j}$  of T for t = d satisfies

$$\lambda_{0,0} = 0, \ \lambda_{0,k} < 0 \quad \forall k = 1, \dots, d..$$

If **w** is a trivial root of  $p_d$ , then it satisfies  $w_{0,k} = 0$  for all k = 1, ..., d. Then, diag(0)(**w**) < 0; this is a contradiction because **w** is supposed to be a root of diag(0).

Let  $d \in \mathbb{N}_{\geq 15}$  be even and  $\mathbf{w}$  be a root of  $\Psi$ . The realization  $p_d = \sum \lambda_{i,j} x_{i,j}$  satisfies  $\lambda_{i,d-i} \neq 0$  if and only if  $i \in \{0,d\}$ . If  $\mathbf{w}$  is a trivial solution of  $p_d$ , the it satisfies  $w_{d,0} > 0$  since it is a root of diag(d). However,  $\operatorname{col}(d-1)(\mathbf{w}) < 0$ , which is a contradiction since  $\mathbf{w}$  is a root of  $\operatorname{col}(d-1)$ .

**Proposition 10.27.** Let T = diag(1) - diag(d-1). Then, T is fixed-contractable.

*Proof.* Use Proposition 10.11.  $\Box$ 

We redefine the following sets.

**Definition 10.28.** We define the following three solution sets:

- (1) Define  $\Gamma^{\text{even}}$  to be the set of all valid contracted hyperfield configurations  $\mathbf{s} \in H^{\Xi}$  such that  $\hat{p}^{\text{even}}(\mathbf{s}) = H$  for all  $p \in \Psi$ .
- (2) Define  $\Gamma^{\text{odd}}$  to be the set of all valid contracted hyperfield configurations  $\mathbf{s} \in H^{\Xi}$  such that  $\hat{p}^{\text{odd}}(\mathbf{s}) = H$  for all  $p \in \Psi$ .

#### **Definition 10.29.** We define

$$\Gamma_6^{\mathrm{even}} \coloneqq \Gamma^{\mathrm{even}} \cap \left\{ \mathbf{s} \in H^\Xi : |\mathrm{supp}^+(\mathbf{s})| = 6 \right\} \text{ and } \Gamma_6^{\mathrm{odd}} \coloneqq \Gamma^{\mathrm{odd}} \cap \left\{ \mathbf{s} \in H^\Xi : |\mathrm{supp}^+(\mathbf{s})| = 6 \right\}.$$

**Proposition 10.30.** All valid outcomes  $\mathbf{w} \in \mathbb{Z}^{V_d}$  with positive support size six satisfy  $\operatorname{contr}_d(\operatorname{sign}(\mathbf{w})) \in \Gamma_6^{\operatorname{even}} \cup \Gamma_6^{\operatorname{odd}}$ .

*Proof.* This follows easily since outcomes are characterized as roots of Pascal forms.  $\Box$ 

**Proposition 10.31.** The following holds:

- (1) We have  $|\Gamma_6^{\text{even}}| = 106806$ .
- (2) We have  $|\Gamma_6^{\text{odd}}| = 110272$ .

We excluded around 100,000 cases; there are still around 217,000 cases left to check.

# 10.3 Reducing Cases with Fixed-Contractable Pascal Forms

The final reduction step relies on fixed-contractable forms. Starting with a set of formal linear combinations of row, col, and diag, our goal is to extract a subset that consists only of fixed-contractable forms. To achieve this, we fix a degree  $D \in \mathbb{N}$ , realize each form at degree D, and algorithmically verify that it is fixed-contractable. Propositions 10.17 and 10.18 provide examples of such algorithmic proofs. The corresponding source code is available on GITHUB TODO. Using this method, we computed a total of 18,273 fixed-contractable forms.

Once a set G' containing only fixed-contractable forms is obtained, we proceed as follows: For each  $\mathbf{s} \in \Gamma_6^{\text{even}}$ , we check whether  $0 \in \text{contr}(p_D)(\mathbf{s})$  for all  $T \in G'$ . If there exists any realization  $p_D$  of T such that  $\mathbf{s}$  is not a hyperfield root, then  $\mathbf{s}$  can be excluded. This exclusion is valid because T is a linear combination of Pascal forms, and all valid outcomes on  $\mathbb{Z}^{V_d}$  must be roots of every Pascal form.

By applying this method, we were able to reduce the number of cases from 106,806 to just 6,700. TODO COMPUTE.

# Chapter 11

# Discussion

This thesis establishes a connection between the classification of discrete statistical models (Theorem 2.23) and a combinatorial puzzle related to chipsplitting games (Theorem 3.32). Specifically, the puzzle investigates whether the degree of a valid chipsplitting outcome can grow indefinitely while its support size remains fixed. For outcomes with positive support sizes up to five, we prove that the degree cannot grow indefinitely, providing a definitive negative answer.

For outcomes with a positive support size of six, significant progress was made toward a similar conclusion. By employing systematic reductions, the number of cases requiring analysis was reduced from approximately 300,000 to 12,000, indicating that a negative answer may hold for this case as well. Furthermore, computational evidence strengthens this hypothesis; our calculations in Table 11.1 reveal that for positive support size n=6, there are no valid outcomes of degree  $d \in \{10, \ldots, TODO\}$ . This pattern suggests that the degree of valid outcomes with positive support size six is bounded above by nine.

$n \setminus d$	1	2	3	4	5	6	7	8	9	10	11
2	1										
3		3	1								
4			12	4	2						
5				82	38	10	4				
6					602	254	88	24	2		
7						6710	2421	643	198	32	4

Figure 11.1: The number of fundamental outcomes for each positive support size n and degree d. For n = 2, 3, 4, 5 there are no additional columns beyond those shown as we have proven that the degree is bounded.

To construct Table 11.1, we employed the following methodology. First, we calculated the set of all supports of valid outcomes for a fixed degree  $d \in \mathbb{N}$  using Algorithm 5. For each chipsplitting support generated, we mapped it back to a statistical model and solved the corresponding linear system to determine whether it yields a fundamental model, as defined in Definition 2.14. The full implementation details and code are available in the repository TODO.

By using this approach, we computed all valid outcomes for positive support sizes  $n = 1, \ldots, 7$  and degree  $d \leq 11$ . This extends the results of Bik and Marigliano [2] by one additional support size, made possible by a more efficient implementation.

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