Any finite number divided by infinity is as near to nothing as makes no odds, so the average population of all the planets in the Universe can be said to be zero. From this it follows that the population of the whole Universe is also zero, and that any people you meet from time to time are merely the products of a deranged imagination.

Douglas Adams (1980, Ch. 19)

Only God can make random selections.

Marion J. Levy (1981, Ninth Law)

2.1 Introduction: from the finite to the transfinite

In this chapter, we deal with the foundations of probability theory inspired by the case of a fair lottery on the natural numbers. Although such a lottery may not objectively exist, we have strong intuitions about it. We may even give quantitative answers to questions such as: “What is the probability that the winning number is odd?” We may answer “Fifty percent”, by considering the probability of even and odd numbers in finite lotteries.

When forming an image of infinite mathematical objects, we rely on our experience with finite objects (Lavine, 1995). More often than not it is impossible to construct or discover an infinite counterpart of a finite concept that fulfils all our intuitions concerning the former. In such cases we have to choose which of those intuitions is most dear to us and weaken or give up at least one other. An example from classical mathematics would be the assignment of cardinalities to infinite sets by Cantor: he took the existence of a one-to-one correspondence between sets as the guiding principle for assigning equal sizes to them, but had to give up the intuition that the whole is always larger than the parts. If the infinite was like the finite in
Chapter 2. Fair Infinite Lotteries

every relevant respect, then it would not be so interesting. Giving up some of our intuitions and tacit assumptions is just the price we have to pay if we want to study a new object. So it is with infinite lotteries. We will have to give up some of our intuitions governing finite lotteries. The question is: which ones, and to what extent.

In this chapter, we consider a fair lottery in which exactly one winner is randomly selected from a countably (or denumerably) infinite set of tickets. We intend to find a description of such a lottery that is mathematically sound and philosophically adequate, by examining our intuitions governing finite lotteries. It may be objected that since there are no infinite lotteries in reality, it is not clear how we can have any intuitions about the concept. Real world lotteries are always finite, but —no matter how the drawing is realized— the outcome can never be guaranteed to be random. Therefore even the idea of a truly fair \( n \)-ticket lottery, where \( n \) is some finite natural number, is a highly idealized concept, but one that is useful in analyzing a broad range of practical situations. Allowing the lottery to have an infinite number of possible outcomes is an additional idealization. The idea of a fair lottery on the natural numbers occurs in probabilistic number theory (Tenenbaum, 1995) and may be a useful approximation for large lottery-like phenomena. This problem also goes by the name of ‘de Finetti’s lottery’ (Bartha, 2004) and ‘God’s lottery’ (McCall and Armstrong, 1989). Although we will never be confronted with a lottery consisting of an infinite set of tickets in reality, it is valid to ask what probability can rationally be assigned to a ticket.

In the subjectivistic approach to probability, it has been argued (for instance by Ramsey, 1931, de Finetti, 1974) that our subjective probability assignments can only be rational if they agree with Kolmogorov’s laws of probability (Kolmogorov, 1933). Within Kolmogorov’s axiomatization however, there simply is no description available for a fair countably infinite lottery. To describe this case, we have to formulate new axioms or at least change one of the assumptions or axioms of Kolmogorov’s system. One solution, advocated by de Finetti (1974), is to relax the requirement of countable additivity to finite additivity. In this chapter, we will develop a different approach. We propose to replace the co-domain (or range) of the probability measure by a non-standard set: this allows us to assign a non-zero, infinitesimal probability to single tickets. In contrast to de Finetti’s solution, there will be a sense in which the probability of a countably infinite union of events supervenes on the probabilities of the individual events. This sense is captured by an additivity principle that is a close analogue of the usual assumption of countable additivity.

The chapter is structured as follows. In section 2.2, we examine our intuitions concerning finite and infinite lotteries. In section 2.3 we review asymptotic density and a generalization thereof by means of which a finitely additive, real-valued probability measure can be obtained on the full power set of the natural numbers. In section 2.4, we introduce some central concepts of non-standard analysis. In section 2.5, we construct a hyperrational-valued probability measure (based on the concept of numerosity) and show that it is hypercountably additive rather than countably additive. In section 2.6, we compare our hyperrational approach with the real-valued solution based on (generalized) asymptotic density, and with a hyperfinite lottery.
In the 2.7th and last section, we review the most salient features of the proposed hyperrational description of a fair, countably infinite lottery.

Regarding notation, throughout this chapter we take \( \mathbb{N} \) to be the set of strictly positive integers, \( \{1, 2, 3, \ldots \} \). \( \text{Even} \) will denote the set of even natural numbers and \( \text{Odd} \) that of the odd natural numbers. We use \( \{ \} \)-brackets to indicate an \( \omega \)-sequence; if only one element is given between the brackets, it will be the general element at a position \( n \in \mathbb{N} \). We abbreviate ‘non-standard analysis’ as NSA, ‘finite additivity’ and ‘finitely additive’ with FA, and ‘countable additivity’ and ‘countably additive’ with CA. The ‘H’ in HFA and HCA adds the prefix ‘hyper-’ to the former abbreviations.

### 2.2 Intuitions concerning lotteries

First we need to be precise about what we mean exactly with a finite lottery, give a mathematical description of it, and make our intuitions about it explicit. Subsequently we investigate to what extent these intuitions carry over to the infinite case, and what has to be changed in the mathematical description to maximize the intuitive appeal of it.

#### 2.2.1 Finite lotteries

**2.2.1.1 Probability measure**

By a finite lottery we mean a process that assigns exactly one winner among a discrete set of tickets in a fair way. By fair we mean that each ticket initially has the same probability of winning. So this process can be modeled by a uniform, discrete function (given below in eq. 2.1), which fulfills all of Kolmogorov’s axioms for probability measures (Kolmogorov, 1933).

The sample space is the set of tickets. The tickets may be numbered, but they need not be: they may be characterized by other symbols with no apparent order. Because the tickets are finite in number, say \( n \in \mathbb{N} \), they can be labeled with an initial segment of the natural numbers, and this set of numbers \( \{1, \ldots, n\} \) may be used as the sample space instead.

The event space is a \( \sigma \)-algebra which contains all combinations of tickets to which we can assign a probability. Since we may do so for any possible combination of tickets, the event space is the powerset \( \mathcal{P} \( \{1, \ldots, n\} \) \) of the sample space. The probability values for an \( n \)-ticket lottery form the set \( \{0, \frac{1}{n}, \frac{2}{n}, \ldots, \frac{n-1}{n}, 1\} \). The co-domain of a probability measure on a finite lottery with an unspecified number of tickets is therefore the set of all rational numbers in the \( [0, 1] \) interval: \( [0, 1]_\mathbb{Q} = [0, 1] \cap \mathbb{Q} \). The probability measure is given by the function:

\[ P(\{1, \ldots, n\}) = \frac{1}{n} \]

---

1 An algebra is a family of subsets of the sample space that contains the sample space itself and is closed under complementation and finite unions; a \( \sigma \)-algebra is closed under denumerable unions on top of that.
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\[ P_n : \mathcal{P}(\{1, \ldots, n\}) \rightarrow [0,1]_\mathbb{Q} \]

where \( \# \) is the counting function, that maps a finite set to its number of elements (finite cardinality). Thus, \( P_n \) is a counting measure normalized by the total number of elements in the sample space. \( \# \) is the prototype of a CA measure, a property that we will employ in the proof of the additivity of the non-standard measure that we will construct.

Thus, \( P_n \) is CA too, but only in a trivial sense: for each countable family of disjoint subsets \( \{A_1, A_2, A_3, \ldots\} \) of the sample space, there will be a finite value \( k \in \mathbb{N} \) such that for all \( m \geq k, \ A_m \) is an empty set. Hence, in the countable sum \( \sum_{m \in \mathbb{N}} P_n(A_m) \), all terms with \( m \geq k \) will be zero and CA reduces to FA in this case.\(^2\)

2.2.1.2 Intuitions

Now we list our intuitions governing a finite lottery. Some of these may seem highly related, and they are—at least in the finite case—, but they need not be in the infinite case, so we name them separately:

**FAIR** The lottery is fair.

**ALL** Every ticket has a probability of winning.

**SUM** The probability of a combination of tickets can be found by summing the individual probabilities.

**LABEL** The labelling of the tickets is neutral with respect to the outcome.

The assumption FAIR embodies the thought that one ticket does not have a higher probability than any other one: a fair lottery is governed by equiprobability. This can only be implemented by the formal requirement that the associated probability function is uniform.

The assumption ALL can only be implemented by the requirement that the probability of any possible combination of tickets is defined. In other words, the probability function must be defined over the whole power set of the event space.

The assumption LABEL is motivated by the intuition that labelling is no more than a convention inspired by the need for referring to specific tickets. It is implemented by requiring of the associated probability function that it is invariant under permutations of the domain.

The assumption SUM is motivated by the intuition that the probability of a set containing the winning number supervenes on the chances of winning that accrue to the individual tickets. The usual assumption of countable additivity (CA, sometimes also called \( \sigma \)-additivity) is one attempt of making the intuition that is encapsulated by SUM precise. We will argue, however, that this is not the right way to do it in

\(^2k \) cannot be fixed in general, but depends on the family. Although at most \( n \) sets can be non-empty, \( k \) may need to be larger than \( n \), since a family of sets can consist of many empty sets ‘in between’ the non-empty ones.
2.2. Intuitions concerning lotteries

In this case. In other words, we will argue that the implementation of SUM is not as straightforward an affair as is commonly thought.

The constraints FAIR, ALL, and SUM jointly entail that every point event must be assigned a non-zero probability. Thus in the context of infinite lotteries the so-called principle of Regularity holds. This is not to say, however, that this principle must hold in all probabilistic scenarios.\(^3\)

These assumptions motivate the standard account of finite lotteries, and they are jointly satisfied by the standard description given by \(P_n\). Let us now briefly survey how these assumptions fare in the context of infinite lotteries.

2.2.2 Infinite lotteries

The infinite counterpart of a finite lottery that we are interested in here is an infinite, denumerable lottery, in particular a lottery that has \(\mathbb{N}\) as its sample space.

For a start, suppose that we are very keen on the intuitions ALL and SUM, and that SUM is formally cashed out as CA. There are (uncountably) many probability distributions that satisfy these two constraints. However, it is easy to see that all of them violate FAIR. But the assumption FAIR simply is non-negotiable. The intuition of fairness is absolutely central to our concept of a lottery. Whereas real world lotteries may never be completely fair, we are considering ideal lotteries. Indeed, when one considers infinite lotteries at all, one is leaving the real world behind anyway.

The assumption LABEL might seem reasonable at first blush, but as a consequence of Cantor’s theory of infinite cardinalities, it will have to be abandoned. Every infinite subset of \(\mathbb{N}\) is in one-to-one onto correspondence with every other infinite subset of \(\mathbb{N}\). So if we insist on invariance under permutation, then every infinite subset of \(\mathbb{N}\) will receive the same probability. This immediately leads to a contradiction with the laws of probability.\(^4\) Thus, whereas for finite sample spaces the labelling of the point events is immaterial, in the infinite case it is of the essence. Giving up LABEL admittedly gives rise to a feeling of discomfit. But if we have a naked choice between giving up some of the laws of probability on the one hand, and giving up LABEL on the other hand, then we should surely take the second option.\(^5\)

The assumption ALL seems negotiable to some extent. For one thing, it is a well-known consequence of the Axiom of Choice that there is no probability measure on the whole of \(P(\mathbb{R})\) (Truss, 1997, Chapter 11). If there are no probability measures on \(P(\mathbb{R})\), then it should perhaps come as no surprise that there are no satisfactory probability measures on \(P(\mathbb{N})\) in the context of infinite lotteries. In any event, we will require of any solution to the infinite lottery problem that to the extent that we have strong intuitions about probabilities of a subset \(A \in \mathbb{N}\), the solution takes

\(^3\)Skyrms (1980) and Lewis defend Regularity in general, whereas Williamson (2007), Hájek (2010) and Easwaran (2010) argue that it cannot hold in all cases (even when infinitesimal probability values are allowed).

\(^4\)By considering Even and Odd on the one one hand, and their union \(\mathbb{N}\) on the other hand, we see that all three sets have the same measure, so FA fails.

\(^5\)For a discussion of the role of LABEL in assigning probabilities in the context of the realism debate, see (Douven et al., 2010).
the probability of $A$ to be defined and in agreement with our intuitions—or else the solution will have to contain a winning story about why our intuitions are mistaken.

Like ALL, the assumption SUM is not so easy to assess. It is well-known that there is no uniform probability function on $\mathcal{P}(\mathbb{N})$ that is CA (de Finetti, 1974). So if we want to preserve FAIR and ALL, and insist on SUM, then we must make it precise in a way that is different from CA.

### 2.3 Asymptotic density: real-valued probability with finite additivity

#### 2.3.1 Limiting relative frequency

Although all infinite subsets of the natural numbers have the same cardinality, there are ways to discriminate the ‘size’ of Even from that of $\mathbb{N}$ for instance. In number theory, the asymptotic density $ad$ (or natural density) of a subset $A$ of $\mathbb{N}$ is defined as follows (e.g. in Tenenbaum, 1995, p. 270):

$$ad(A) = \lim_{n \to \infty} \frac{\#(A \cap \{1, \ldots, n\})}{n} \quad \text{if the limit exists} \quad (2.2)$$

Asymptotic density captures the idea that a lottery over $\mathbb{N}$ is obtained from a finite lottery (eq. 2.1) in the limit of the number of tickets, $n$, going to infinity. Thus, for a set $A$ that has a defined asymptotic density $ad(A) = \lim_{n \to \infty} P_n(A)$. Because $\mathbb{Q}$ is not closed under the limit-operation, the co-domain of $ad$ is the $[0, 1]$-interval of the real numbers, rather than $[0, 1]_\mathbb{Q}$ as in the finite case.

Asymptotic density gives rise to a FAIR probability assignment. However, it fails ALL: since asymptotic density is not defined for all subsets of $\mathbb{N}$, it cannot be introduced as a measure with $\mathcal{P}(\mathbb{N})$ as its domain (Tenenbaum, 1995). The collection of all subsets of $\mathbb{N}$ that do possess asymptotic density is not closed under intersection and union, so it does not form an algebra.

#### 2.3.2 A generalization

It is possible to extend asymptotic density to a measure that assigns a value to all subsets of $\mathbb{N}$ (Schurz and Leitgeb, 2008). This requires a generalization of the limit-concept, that assigns a value to all bounded —convergent and non-convergent— sequences: the Hahn-Banach limit (HB-lim) is a real-valued generalization, that equals the value of the classical limit for convergent sequences.\(^7\) For any $A \in \mathcal{P}(\mathbb{N})$ the sequence $\left\{ \frac{\#(A \cap \{1, \ldots, n\})}{n} \right\}$ is bounded (by 1). Therefore, the Hahn-Banach limit of this sequence is defined on all of $\mathcal{P}(\mathbb{N})$, giving rise to the probability measure $P_{ad}$:

---

\(^6\) $ad$ is defined for all finite subsets of $\mathbb{N}$, but not for all infinite subsets. An example of a set for which $ad$ is undefined is the set of numbers whose binary notation contains an even number of digits. A second example comes for free: consider its complement, the set of numbers whose binary representation contains an odd number of digits.

\(^7\) We will give a definition of HB-lim in the non-standard framework in paragraph 2.4.5.
P_{ad} : \mathcal{P}({1, \ldots, n}) \rightarrow [0, 1]_\mathbb{R} \\
A \mapsto \text{HB-lim} \frac{\#(A \cap \{1, \ldots, n\})}{n} \quad (2.3)

The great selling point of this construction is that probabilities are aligned with limiting relative frequencies whenever these are defined. But again, the resulting measure is only FA (Schurz and Leitgeb, 2008). So even though ALL can be obtained, SUM does not hold.

P_{ad} can be thought of as giving precise content to a suggestion by de Finetti: by weakening the requirement for CA to FA, all other axioms of classical probability theory can be saved. This solution to the problem of the fair infinite lottery was advocated by de Finetti (1974). It allows us to assign probability values to infinite subsets of \( \mathbb{N} \) that correspond well to our intuitions (such as a probability 1/2 for Even as well as for Odd). The solution does come with a major drawback, however. It amounts to giving up on the intuition that the chance of a ticket from an infinite set winning is an infinite sum over the chances of individual tickets from the set winning (SUM). As a result, regularity fails, so we also have to give up on the intuition that each ticket has a non-zero chance of winning.

2.4 Infinitesimals

With FA, we can save FAIR, ALL and a finite version of SUM. But it seems odd that the measure of singletons can be exactly 0 while the measure of their union is 1. Now the question is: can we do better? So far, we have only looked at real-valued probability functions, and we have seen that in that framework the answer is ‘no’. Now we will reconsider the question. It seems that we can do better indeed, by assigning an infinitesimal probability to the singletons, rather than 0. In the finite case, the probability of a singleton is 1 divided by the number of elements in the sample space. Since our sample space is \( \mathbb{N} \), with an infinite number of elements, we should assign the inverse of an infinite number to the singletons. Cantor’s cardinalities are not suitable for this, since they have no inverse. Non-standard analysis provides a consistent way of working with unbounded or infinite numbers, which do have an inverse: infinitesimals. So, non-standard analysis allows us to have a different co-domain for the probability measure: \( {}^*\mathbb{R} \), which is the unit interval within \( {}^*\mathbb{Q} \), a non-standard extension of \( \mathbb{Q} \).

We will give a short overview of some essential concepts in non-standard analysis (NSA) (Robinson, 1966). We do not aim at completeness here, but restrict our attention to those ingredients that we will need in the course of this chapter. The information of this section is mainly based on (Cutland, 1983) and Benci et al. (2006a).

2.4.1 The star-map and Transfer

All approaches to NSA need a tool that maps any standard object, \( A \) (which can be a number, set, function, \ldots), to its unique non-standard counterpart or
hyperextension, \( *A \). This function, called the star-map, should preserve a large class of properties, which is ensured by the Transfer principle. In axiomatic approaches to NSA, the Transfer principle is stipulated as an axiom; in other approaches such as the ultrafilter-construction that we will employ here, it is a theorem.

An important example is the star-map of an \( \omega \)-sequence. This is a hypersequence, \textit{i.e.} a sequence that takes values on all of \( \mathbb{N} \), which is a nonstandard extension of \( \mathbb{N} \). We introduce the notation \( \{ \} \) to distinguish a hypersequence from a standard sequence; if only one element is given between the brackets, it will be the general element at position \( N \), with \( N \in \mathbb{N} \). (We will encounter the star-map of a sequence of sets in eq. (2.23).)

The star-map can be obtained using free ultrafilters; this requires the introduction of equivalence classes determined by a free ultrafilter.

### 2.4.2 Equivalence classes based on a free ultrafilter

Consider the set of \( \omega \)-sequences on a general set \( X \) (or the set \( X^\mathbb{N} \) of functions \( \mathbb{N} \to X \)). The idea is to interpret a whole sequence as one entity, be it a non-standard one. Even if \( X \) is a set of numbers, when \( X^\mathbb{N} \) fails to form a field, it does not provide a useful number system. To this end, we need to make a choice of ‘which positions in the sequence matter’. For instance, a difference in only finitely many positions should not matter. Fixing a free ultrafilter on the label set, \( \mathbb{N} \), is a way to settle all these choices at once.\(^8\) A nice introduction to ultrafilters can be found in (Komjáth and Totik, 2008).

A free (or non-principal) ultrafilter, \( \mathcal{U} \), on \( \mathbb{N} \) is a collection of subsets of \( \mathbb{N} \)\( (\mathcal{U} \subset \mathcal{P}(\mathbb{N})) \), which fulfills four requirements:

1. \( \emptyset \notin \mathcal{U} \)
2. \( (\forall A, B \in \mathcal{U}) \ A \cap B \in \mathcal{U} \)
3. \( (\forall A \subset \mathbb{N}) \ A \notin \mathcal{U} \Rightarrow \mathbb{N}\setminus A \in \mathcal{U} \) (ultra)
4. \( (\forall A \subset \mathbb{N}) \ A \) is finite \( \Rightarrow \mathbb{N}\setminus A \in \mathcal{U} \) (free)

Two sequences are equivalent (or equal ‘almost everywhere’) with respect to a free ultrafilter just if the set of labels where their terms are exactly equal is an element of the filter:

\[
(\forall (x_n), (y_n) \in X^\mathbb{N}) \quad (x_n) \approx_{\mathcal{U}} (y_n) \iff \{ n \mid x_n = y_n \} \in \mathcal{U}
\]  

We may define the equivalence class of a sequence \( (x_n) \) modulo the just defined equivalence relation, \( [(x_n)]_{\mathcal{U}} \), as follows:

\(^8\)In general, non-standard analysis may be developed from considering a free ultrafilter on any infinite set, but in this chapter we will always use \( \mathbb{N} \) as the index set.
2.4. Infinitesimals

\[(\forall (x_n) \in X^N) \ [ (x_n) \mathrel{\equiv} \{ (y_n) \in X^N \mid \langle y_n \rangle \mathrel{\equiv} \langle x_n \rangle \} \] (2.5)

The set of equivalence classes of sequences does provide a good basis for a number system and may be interpreted as \(*X\), the hyperextension of \(X\).

2.4.3 \(*N\) and \(*Q\)

To illustrate how the equivalence class of sequences can provide the star-map of a set \(X\), we consider the hyperextensions of \(X = \mathbb{N}\) and \(X = \mathbb{Q}\).

\(*\mathbb{N}\) is defined as the set of equivalence classes (determined by the choice of a free ultrafilter on \(\mathbb{N}\)) of elements of \(\mathbb{N}^N\), \([(m_n)]\mathcal{U}\). The elements of \(*\mathbb{N}\) are called hypernatural numbers. The equivalence class of a constant sequence \([\langle m_n \rangle]\), with \(m_n = m \in \mathbb{N}\) for all \(n \in \mathbb{N}\), can be written in short as \(*m\) and may be identified with the standard natural number \(m\): this embeds \(\mathbb{N}\) in \(*\mathbb{N}\). Because of the construction, \(*\mathbb{N}\) is called a sequential extension of \(\mathbb{N}\): it consists of (equivalence classes of) \(\omega\)-sequences of standard natural numbers.

For probability values, the sequential extension of \(\mathbb{Q}\) seems more appropriate than \(*\mathbb{N}\). The set of hyperrational numbers \(*\mathbb{Q}\) can be obtained in several ways. Because the set of integers, \(\mathbb{Z}\), can be introduced as the closure of \(\mathbb{N}\) under substraction, and the set of rational numbers, \(\mathbb{Q}\), can be introduced as the fraction field of \(\mathbb{Z}\), \(*\mathbb{N}\) can be extended similarly to \(*\mathbb{Z}\) and \(*\mathbb{Q}\) subsequently. Alternatively, by considering sequences of integer or rational numbers, \(*\mathbb{Z}\) and \(*\mathbb{Q}\) can be constructed using ultrafilters in a similar fashion as \(*\mathbb{N}\). (As was already mentioned, the star-map returns the hyperextension of any standard object, so also \(*\mathbb{R}\) and \(*\mathbb{C}\) can be obtained.)

2.4.4 \(\mathbb{R}\) as an approximation to \(*\mathbb{Q}\)

To relate results of NSA to standard analysis, the standard part (or shadow) function, \(st\), is a useful concept: \(st\) maps any non-standard number to the closets real value, which is uniquely determined. Clearly, ‘taking the standard part’ comes down to rounding up to infinitesimals.

For fair lotteries on finite sample spaces, the probabilities are fractions (rational numbers). For infinite sample spaces, probabilities can be associated with \(\omega\)-sequences of rational values. Observe that both the construction of \(\mathbb{R}\) from Cauchy sequences and the above construction of \(*\mathbb{Q}\) start from \(\omega\)-sequences of rational numbers. Only the rule by which the whole sequence is associated with a new —real or hyperrational— number differs. Thus, for the range of a fair infinite lottery we seem to have a choice between \(\mathbb{R}\) and \(*\mathbb{Q}\) (rather than between \(\mathbb{R}\) and \(*\mathbb{R}\)).

If we use the standard part function on \(*\mathbb{Q}\), we can obtain any value of \(\mathbb{R}\). Thus, from the non-standard viewpoint, real-valued probabilities can be seen as an approximation of hyperrational numbers. If the real values give satisfactory answers, they are all we need. If not, we may need to look at a more precise description in terms of hyperrationals.
2.4.5 Limits

2.4.5.1 Alpha-limit

We may call \( *a_\alpha \) the ‘ideal value’ or the ‘alpha-limit’ of the sequence \( (a_n) \). It is equal to the value of the hypersequence \( \langle *a_N \rangle \) at position \( N = \alpha \) and also to the ultrafilter equivalence class of the sequence \( (a_n) \) (Benci and Di Nasso, 2003a, p. 367):

\[
* a_\alpha = [\langle a_n \rangle]_\mathcal{U} \tag{2.6}
\]

2.4.5.2 Classical limit

The definition of the classical limit of a real-valued sequence, \( \lim_{n \to \infty} a_n \), can be reformulated in NSA as follows (Väth, 2007, p. 88): \( (a_n) \) converges with limit \( L \in \mathbb{R} \) if and only if

\[
(\forall N \in \mathbb{N} \setminus \mathbb{N}) \quad *a_N - L \text{ is infinitesimal} \tag{2.7}
\]

In particular for \( N = \alpha \), if \( \lim_{n \to \infty} a_n = L \), then \( st(*a_\alpha) = L \). By equation (2.6), we find that:

\[
\lim_{n \to \infty} a_n = st(\langle a_n \rangle) \quad \text{if the limit exists} \tag{2.8}
\]

2.4.5.3 Hahn-Banach limit

Hahn-Banach limits (HB-lim) are a real-valued generalization of the limit that is defined for all bounded sequences and is equal to the value of the classical limit if the sequence converges. In NSA, the Hahn-Banach limit of a real-valued sequence \( (a_n) \) is defined as follows (Väth, 2007, p. 133): \( (a_n) \) is bounded with HB-lim\( (a_n) = L \in \mathbb{R} \) if and only if

\[
L = st \left( \sum_{N = H_0}^{H_1} h_N * a_N \right) \tag{2.9}
\]

for some \( H_0, H_1 \in \mathbb{N} \setminus \mathbb{N} \) with \( H_0 < H_1 \) and for some internal sequence of hyperreals \( h_{H_0}, \ldots, h_{H_1} \) such that \( \sum_{N = H_0}^{H_1} h_N = 1 \).

2.4.6 Internal and external objects

Not all non-standard objects are the image of some standard object by the star-map; those that are, are called ‘internal’, the others ‘external’. This distinction is important but not easy to understand immediately. As a first example, \( \mathbb{N} \) is external; in fact, every infinite countable set is external. It is also important to note that whereas the star-map preserves most set-operations, it does not do so for the powerset: the operation \( ^*\mathcal{P} \) returns only the internal subsets of a given set, and so, for any infinite standard set \( A \): \( ^*\mathcal{P}(A) \notin \mathcal{P}(^*A) \). The probability function we will construct will also turn out to be external.
2.5 Hyperrational valued probability

Let us now construct a non-standard valued probability measure to describe a lottery on \(\mathbb{N}\). If we want to find a function that is formally similar to the probability measure of a finite lottery, then we must find a way to ‘count’ finite as well as infinite subsets of \(\mathbb{N}\) and divide by the size it assigns to the whole sample space to normalize this function.

The construction proceeds in four steps. (1) Every subset of \(\mathbb{N}\) can be represented as an infinitely long bit string by considering its characteristic function. (2) Then we consider the sequence of partial sums of these bits. (3) By introducing a free ultrafilter on \(\mathbb{N}\), we can interpret the whole partial sum sequence as one nonstandard (or hypernatural) number: its numerosity. (4) By a suitable normalization, we finally obtain a hyperrational-valued probability measure on \(\mathcal{P}(\mathbb{N})\).

After the construction we show that numerosities and probabilities based on them are hypercountably additive (HCA). In fact, we shall see that it is perhaps more appropriate to call them hyperfinitely additive (HFA).

2.5.1 The construction

2.5.1.1 Step 1: Characteristic bit string

First, consider the indicator function or characteristic function of a subset \(A\) of \(\mathbb{N}\): it tests whether a natural number is in the set \(A\) or not, where a positive answer corresponds to 1 and a negative to 0.

\[
\chi_A : \mathbb{N} \to \{0, 1\} \\
n \mapsto \begin{cases} 
0 & \text{if } n \notin \mathbb{N}\setminus A \\
1 & \text{if } n \in A 
\end{cases}
\]  

(2.10)

Now we can construct the function that assigns a characteristic bit string (CBS) to any subset of the natural numbers:

\[
CBS : \mathcal{P}(\mathbb{N}) \to \{0, 1\}^\mathbb{N} \\
A \mapsto (\chi_A(1), \chi_A(2), \ldots, \chi_A(n), \ldots)
\]  

(2.11)

In shorthand notation, we can refer to a sequence by its n-th element only, so \(CBS(A) = (\chi_A(n))\).

2.5.1.2 Step 2: Partial sums of characteristic bit strings

Now we consider the sequence of partial sums of characteristic bit strings of a subset \(A\) of natural numbers, \(\text{SumCBS}(A)\).

\[
\text{SumCBS} : \mathcal{P}(\mathbb{N}) \to \mathbb{N}^\mathbb{N} \\
A \mapsto (S_n)
\]  

(2.12)

with
\[ S_n = \sum_{m=1}^{n} \chi_A(m) \]
\[ = \chi_A(1) + \chi_A(2) + \ldots + \chi_A(n) \]  
(2.13)

So \( S_n \) has a value in \( \{0, \ldots, n\} \) for all \( n \). Alternatively, \( S_n \) can be written in terms of the counting function \( \# \):

\[ S_n = \sum_{m=1}^{n} \chi_A(m) \]  
(2.14)
\[ = \sum_{m\in\mathbb{N}} \chi_{A\cap\{1,\ldots,n\}}(m) \]  
(2.15)
\[ = \#(A\cap\{1,\ldots,n\}) \]  
(2.16)

The sequence \( \langle S_n \rangle \) seems to ‘point to a value at infinity’. If we interpret this sentence in the framework of standard analysis, we should take the limit \( n \to \infty \). For infinite sets \( A \) this results in \( \lim_{n \to \infty} S_n = \infty \), which cannot be normalized. (If we consider the sequence \( \langle \frac{S_n}{n} \rangle \) instead and its limit \( n \to \infty \), we find asymptotic density again.) As we have seen however, in NSA we may alternatively interpret the whole sequence as one non-standard number.

### 2.5.1.3 Step 3: Numerosity as the equivalence class of the partial sum sequence

The equivalence class under a free ultrafilter of a sequence of partial sums of characteristic bit strings is a hypernatural number that can be interpreted as the size of the corresponding set, called its **numerosity**.

\[ \text{num} : \mathcal{P}(\mathbb{N}) \rightarrow {}^*\mathbb{N} \]
\[ A \mapsto [(S_n)_{\mathcal{U}}] \]  
(2.17)

with \( S_n \) as before.

Any finite set has a finite numerosity (value in \( \mathbb{N} \), equal to \( \#(A) \)), whereas any infinite set has an infinite numerosity (value in \( {}^*\mathbb{N}\setminus\mathbb{N} \)). As an example, consider \( A = \mathbb{N} \).

In that case, \( CBS(A) = \{1, 1, 1, \ldots, 1, \ldots\} \) and \( \text{num}(A) = \{1, 2, 3, \ldots, n, \ldots\} \) \( \mathcal{U} \), which is larger than any finite number. Thus we have shown that \( \text{num}(\mathbb{N}) \) is an element of \( {}^*\mathbb{N}\setminus\mathbb{N} \). We may call this new number, \( \text{num}(\mathbb{N}) \), alpha (\( \alpha \)).

Using equation (2.14), we can transform the definition for numerosity given by equation (2.17) in the following way:

\[ \text{num}(A) = [(S_n)_{\mathcal{U}}] \]  
\[ = [(\#(A\cap\{1,\ldots,n\}))]_{\mathcal{U}} \]  
\[ = [{}^*\#((A\cap\{1,\ldots,n\}))]_{\mathcal{U}} \]  
(2.18)

for all subsets \( A \) of \( \mathbb{N} \).

Note that \( [(\{1,\ldots,n\})]_{\mathcal{U}} \) is a hyperfinite set, which means that there is an infinite hypernatural number, \( N \), such that this set is equal to

---

9 The form on the second line makes it clear that our ultrafilter-based definition of the numerosity function is equal to that of the axiomatic approach developed in (Benci and Di Nasso, 2003b). In
2.5. Hyperrational valued probability

{1,\ldots,N}. We can even be more precise, the hypernatural number \( N \) is equal to \([(1,2,3,\ldots,n,\ldots)]_{\mathcal{U}}\), which we will call \( \alpha \) (following the terminology of Benci and Di Nasso). So we obtain:

\[
num(A) = \ast \#(A \cap \{1,\ldots,\alpha\})
\]  

(2.19)

Remark how similar this form is to the numerator of the asymptotic density function in equation (2.2). Thus, equation (2.19) is very suggestive for a probability function: all we need to do is normalize it.

2.5.1.4 Step 4: Non-standard probability

The construction is completed by normalization of the numerosity function. By dividing it by the numerosity of the sample space, \( num(\mathbb{N}) = \alpha \), we can introduce the probability function of a lottery over \( \mathbb{N} \) in a form that is similar to the probability measure of a finite lottery (eq. 2.1):

\[
P_{num} : \mathcal{P}(\mathbb{N}) \rightarrow \ast [0,1]_{\mathbb{Q}} \quad \text{for } \frac{\# A}{num(A)}
\]  

(2.20)

\( P_{num} \) takes values on the unit interval of \( \ast \mathbb{Q} \) and may be interpreted as a hyperrational-valued probability function. It assigns an infinitesimal probability to any finite set and a larger probability to any infinite set.

Since algebraic operations on non-standard numbers are equivalent to the termwise application of the corresponding operation on the underlying sequence, from our construction we obtain:

\[
P_{num}(A) = \frac{\frac{\#(A \cap \{1,\ldots,n\})}{n}}{\alpha} \in [\mathbb{N}]_{\mathcal{U}}
\]  

(2.21)

which makes the analogy with asymptotic density complete.

\( P_{num} \) is an external object, because its domain is \( \mathcal{P}(\mathbb{N}) \), which is an external set. This means that the function cannot be obtained directly by taking a standard function and applying the Transfer Principle to it.

2.5.2 Additivity of the probability function

In this paragraph, we will investigate to what extent our proposal for \( P_{num} \) satisfies SUM. We will show that the numerosity function is hypercountably additive (HCA) rather than countably additive (CA). It then follows directly that the probability measure based on it is HCA too.

the existence proof given in that paper, numerosity is related to the ultrafilter-construction of NSA as follows: \( num(A) = \ast \#(A \cap \{1,\ldots,n\}) \in [\mathbb{N}]_{\mathcal{U}} \) (Benci and Di Nasso, 2003b, p. 62), where \( \mathcal{U} \) is a free ultrafilter such that \( \mathcal{U} = \{ A \subseteq \mathbb{N} \mid \alpha \in \ast A \} \) (Benci and Di Nasso, 2003b, p. 374).
2.5.2.1 Addition on $^*\mathbb{N}$

First we need to define finite addition on $^*\mathbb{N}$. The sum of two hypernatural numbers is defined as the star-map of the standard sum operation (on standard numbers) $+: ^*\mathbb{N} \times ^*\mathbb{N} \to ^*\mathbb{N}$. For any two sequences $\langle a_n \rangle$ and $\langle b_n \rangle$ we have: $\langle (a_n) \rangle_U + \langle (b_n) \rangle_U = \langle (a_n + b_n) \rangle_U$. Since it will be clear from the context whether we are summing standard or non-standard numbers, we may drop the $^*$ from the sum-symbol. Likewise, all finite sums are defined.

The countably infinite sum $\sum_{n \in \mathbb{N}}$ is not defined for non-standard terms: since $\sum_{n \in \mathbb{N}} = \lim_{m \to \infty} \sum_{n=1}^{m}$ and limits are only defined for standard numbers. Another way to see this is by taking into account that $\mathbb{N}$ is an external set, and summations over such sets are not defined.

An infinite addition that is relevant (always defined) for non-standard numbers is the hyperfinite sum: the equivalence class of a sequence of finite sums. We may also consider the summation over all of $^*\mathbb{N}$ —a hypercountable sum $\sum_{N \in ^*\mathbb{N}}$— since the latter is an internal set.

2.5.2.2 $\textit{num}$ is not CA

CA relates a property of the domain of a measure to a property of its range. First of all, the domain has to be a $\sigma$-algebra, ensuring that the union of any countably infinite family of sets in the domain is also in the domain. Secondly, countable sums have to be defined on the range. CA then links the two by requiring that the measure of the union of a countable family is equal to the countable sum of measures of the each of the members of the family.

The domain of the numerosity function is $\mathcal{P}(\mathbb{N})$, which is indeed a $\sigma$-algebra. The range of the function however is the set of hypernatural numbers, for which the countably infinite sum is undefined. Therefore, the numerosity function cannot be countably additive. The same argument applies to $\textit{any}$ function that maps $\mathcal{P}(\mathbb{N})$ to a set of non-standard values.

2.5.2.3 $\textit{num}$ of a sequence of sets

So far, the numerosity function is only defined for individual subsets of $\mathbb{N}$ (eq. 2.17). Now, we define the numerosity of a sequence of disjoint subsets as follows:

$$\text{num}(\langle A_n \rangle) = ^*\#(\langle (A_n) \rangle \cap \{1, \ldots, \alpha\})$$  \hspace{1cm} (2.22)

The goal of this paragraph is to find an equivalent form of the above definition, that gives us more insight in the additivity of this function.

First, consider a sequence of (possibly empty) subsets of $\mathbb{N}$:

$$\langle A_1, A_2, \ldots, A_n, \ldots \rangle = \langle A_n \rangle_{n \in \mathbb{N}}$$

The star-map of this sequence is a hypersequence of internal subsets of $^*\mathbb{N}$:

$$\langle (A_n)_{n \in \mathbb{N}} \rangle = \langle ^*A_N \rangle_{N \in ^*\mathbb{N}}$$  \hspace{1cm} (2.23)
with \( *A_N = A_N \) for \( N \in \mathbb{N} \).

Define the intersection with and the hypercounting function of a hypersequence of subsets of \( {}^*\mathbb{N} \) componentwise:

\[
\forall S \in {}^*(\mathcal{P}(\mathbb{N})) \quad \langle {}^*A_N \rangle \cap S = \langle {}^*A_N \cap S \rangle \tag{2.24}
\]

\[
*\#(\langle {}^*A_N \rangle) = \langle *\#({}^*A_N) \rangle \tag{2.25}
\]

By applying equations 2.23, 2.24 and 2.25, the definition given in equation (2.22) is transformed to:

\[
um((A_n)) = \langle *\#({}^*A_N \cap \{1, \ldots, \alpha\}) \rangle \tag{2.26}
\]

Thus, the numerosity of an \( \omega \)-sequence of sets is a hypersequence of hypernatural numbers; we may refer to its \( N \)-th element as \( (\num((A_n)))_N \).

### 2.5.2.4 \textit{num} and \( P_{num} \) are HCA

Now we arrive at the main result of this chapter: the numerosity function is hypercountably additive. To prove this, we need to show that for any family of disjoint subsets of \( \mathbb{N} \), \( \{A_n \mid n \in \mathbb{N}\} \):

\[
um(\bigcup_{n \in \mathbb{N}} A_n) = \sum_{N \in {}^*\mathbb{N}} \left( \num((A_n)) \right)_N \tag{2.27}
\]

As an essential step in the proof, we will use that the hyperextension of a countable union of a countable family of sets is equal to the hypercountable union of the hyperextension of the family (by Transfer, cf. Rubio, 1994, p. 104):

\[
*\left( \bigcup_{n \in \mathbb{N}} A_n \right) = \bigcup_{N \in \mathbb{N}} *A_N \tag{2.28}
\]
Proof.

\[
\text{num}(\bigcup_{n \in \mathbb{N}} A_n) = \ast \# \left( \ast \left( \bigcup_{n \in \mathbb{N}} A_n \right) \cap \{1, \ldots, \alpha\} \right)
\]

[By definition of num in equation (2.19)]

\[
= \ast \# \left( \bigcup_{N \in \ast \mathbb{N}} \ast A_N \cap \{1, \ldots, \alpha\} \right)
\]

[By equation (2.28)]

\[
= \ast \# \left( \bigcup_{N \in \ast \mathbb{N}} (\ast A_N \cap \{1, \ldots, \alpha\}) \right)
\]

[Distributivity of intersection over union + Transfer]

\[
= \sum_{N \in \ast \mathbb{N}} \ast \#(\ast A_N \cap \{1, \ldots, \alpha\})
\]

[CA of counting function + Transfer]

\[
= \sum_{N \in \ast \mathbb{N}} (\text{num}(A_n))_N
\]

[By equation (2.26) + remark below it]

\[\square\]

Because the numerosity function is HCA, so is the probability function obtained from it (provided that we first define \(P_{\text{num}}\) of a sequence of sets, as a normalized version of eq. 2.22). In particular, for the entire sample space \(\mathbb{N}\), the infinitesimal probabilities of the countably infinite family of singletons do add up to unity.

The idea behind the proof is visualized in Figure 2.1 for the specific case of the countably infinite family of singletons of \(\mathbb{N}\). In this example, we consider the countably infinite family of singletons of \(\mathbb{N}\), whose union is \(\mathbb{N}\). The numerosity function of a single set requires the extension of sequences (SumCBS) into hypercountably long ones (\(\ast\)SumCBS) and their validation at position alpha. To determine the numerosity of a countably infinite union by addition, the countable family of standard sets has to be extended to a hypercountable family of non-standard sets. By looking at the last horizontal line, we see that the numerosity of \(\bigcup_{n \in \mathbb{N}} A_n\) (here \(\bigcup_{n \in \mathbb{N}}\{n\} = \mathbb{N}\)) is determined as \(\ast \#(\ast \left( \bigcup_{n \in \mathbb{N}} A_n \right) \cap \{1, \ldots, \alpha\})\), here \(\ast \#(\ast \mathbb{N} \cap \{1, \ldots, \alpha\}) = \oplus\). In this case, the values in the column with number \(M = \alpha\) add up to \(\oplus\) too. For each case, a similar table can be made, and the proof states that the same value is always obtained from comparing the \(\alpha\)th position of the last row with the hypercountable sum of the \(\alpha\)th column.

It is apparent from the example in Figure 2.1 that the lottery on \(\mathbb{N}\) is HCA in a very specific sense: for each family, there can always be found a hypernatural number, \(K \in \ast \mathbb{N}\) (\(\alpha\) in the example, but possibly larger in other cases), such that the hypercountable sum decomposes in a hyperfinite sum and a hypercountable tail with zero-terms only. Thus we may call \(\text{num}\) and \(P_{\text{num}}\) hyperfinitely additive (HFA).

\[\text{10}\] This is completely analogous to the argument, given in section 2.2.1, showing that a finite lottery is only CA in a trivial sense.
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Figure 2.1: Illustration of the proof for the hypercountable additivity of the numerosity function for the specific case of the countably infinite family of singletons of \( \mathbb{N} \). (Further explanation can be found in the main text.)
2.6 Discussion

In this section, first we evaluate some consequences of our use of free ultrafilters to construct generalized probability functions for infinite lotteries. Subsequently we investigate the relation of a lottery on \( \mathbb{N} \) to a different type of infinite lottery: the hyperfinite case. We also compare the non-standard description of a lottery on \( \mathbb{N} \) to the best available real-valued approach, that of generalized asymptotic density; we see that SUM is lost in the latter as a result of an accumulation of rounding errors.

2.6.1 Non-constructiveness

Our approach may be criticized on mildly constructivist grounds. The existence proof for free ultrafilters uses Zorn’s lemma and thus depends on the acceptance of the Axiom of Choice. Thus a hyperrational-valued probability measure requires free ultrafilters, which depend on the Axiom of Choice (AC).

To this objection we may reply that the more common solution based on FA, asymptotic density (paragraph 2.3) requires an extension of the limit concept measure that involves a free ultrafilter too. Therefore, this generally accepted solution is non-constructive too, as has been pointed out by Lauwers. To those who are unwilling to accept the Axiom of Choice, no measure is available that does any justice to the intuitions underlying fair lotteries. To the rest of us who have no objection to the Axiom of Choice, the HCA, hyperrational probability function is no less acceptable than the FA real-valued one. But there is something to be said in favour of the hyperrational probability function: it gives the SUM-intuition its due.

2.6.2 Non-uniqueness

Elga (2004) has remarked that there are often too many non-standard solutions: if we can use any infinite hypernatural number to model a problem, why should we prefer one rather than another? One answer is given by alpha-theory, which develops NSA from the idea of adding a new ideal number, \( \alpha \), to \( \mathbb{N} \). This \( \alpha \) can be interpreted as the numerosity of the set \( \mathbb{N} \) (Benci and Di Nasso, 2003a, p. 357). Since numerosity theory introduces \( \alpha \) as the size of the natural numbers, and constructs the other hypernatural numbers around it, \( \alpha \) has a clear interpretation that is hardwired into the theory. It gives us a point of reference among the infinite number of infinite hypernaturals.

We have started from the ultrafilter-construction of \( \ast \mathbb{N} \) and then considered the equivalence class of the sequence \([\{1,2,3,\ldots\}]_U\), which we interpret as \( \alpha \) (consistent with the axiomatic approach). So, in the present context the accusation of arbitrariness boils down to the choice of a free ultrafilter \( U \). A different choice of free ultrafilter produces a different value of \( \alpha \) and hence a probability function with the same standard part but infinitesimal differences.

To come back to the Even versus Odd example: the odd tickets always have a head-start compared to the even ones, for the simple reason that 1 comes before 2. Within our framework, it should not come as a surprise that the weight of this
very first ticket may result in an infinitesimal advantage for the whole set of odd tickets. After all, our approach is based on the idea that even in an infinite lottery each ticket has a non-zero probability. For all finite cases with an odd number of tickets, \( P_n(\text{Odd}) = \frac{1}{2} + \frac{1}{2n} > \frac{1}{2} \) and \( P_n(\text{Even}) = \frac{1}{2} - \frac{1}{2n} < \frac{1}{2} \), whereas \( P_n(\text{Odd}) \) and \( P_n(\text{Even}) \) are exactly \( \frac{1}{2} \) for all finite lotteries with \( n \) even. Because the probability assignment in both cases is different for all finite lotteries, it leads to two different infinite lotteries as well: one in which the odd numbers are in the ultrafilter, such that \( P_{\text{num}}(\text{Odd}) = \frac{1}{2} + \frac{1}{2\alpha} \) and \( P_{\text{num}}(\text{Even}) = \frac{1}{2} - \frac{1}{2\alpha} \), and one in which the even numbers are in it, such that \( P_{\text{num}}(\text{Odd}) = \frac{1}{2} \) and \( P_{\text{num}}(\text{Even}) = \frac{1}{2} \).

It may look as if we can favor the second solution: a good probability measure should at least respect the limiting frequencies, and they are exactly \( \frac{1}{2} \) for \text{Even} and \text{Odd}. Take into account however, that limiting frequencies are real-valued, and that our two non-standard solutions have exactly the same standard part, so they are both in accordance with this \( \frac{1}{2} \) \& \( \frac{1}{2} \) solution: limiting frequencies do not rule out one of both.

By observing that \( \text{Even} = \{ n \in \mathbb{N} \mid n \mod 2 = 0 \} \) and \( \text{Odd} = \{ n \in \mathbb{N} \mid n \mod 2 = 1 \} \), the above analysis generalizes: for each \( m \in \mathbb{N} \), \( m \) different scenario’s emerge (corresponding to the issue whether \( \{ n \in \mathbb{N} \mid n \mod m = k \} \) is in the ultrafilter for either \( 0, 1, \ldots \), or \( m - 1 \)).

It is possible to impose additional constraints on the ultrafilter, such that the set corresponding to \( m \) mod 0 is in the ultrafilter for all \( m \in \mathbb{N} \), or equivalently: such that \( \alpha \) is a multiple of any finite number (Benci and Di Nasso, 2003b, Mancosu, 2009). However, we see at present no convincing reasons for endorsing any particular constraints of this kind. We already remarked that the infinite lottery violates LA-BEL. Here we encounter another difference with finite lotteries: the solution is not unique. The problem stated as “Find a probability measure on all of \( \mathbb{N} \) that satisfies FAIR, ALL and SUM” is highly underdetermined: there are as many different ways to draw a random number from \( \mathbb{N} \) in a fair way as there are free ultrafilters, and the probability function \( P_{\text{num}} \) given in equation (2.20) should be seen as a whole family of solutions, all of which are, as far as we can presently see, equally relevant.

### 2.6.3 Lotteries on \( \mathbb{N} \) versus hyperfinite lotteries

In non-standard measure theory, the emphasis lies on internal measures: measures that can be obtained as the star-map of a standard function. Since \( \mathbb{N} \) is an external set within \( ^*\mathbb{N} \), no internal measure is appropriate to describe the probabilities concerning a lottery over \( \mathbb{N} \). The probability measure we propose in this chapter is indeed an external function, but it is closely related to an internal one: the probability function of a hyperfinite lottery.

Consider the finite set \( A_n = \{1, \ldots, n\} \). Then \( \{A_n\}_\mu = \{1, \ldots, \alpha\} \) is a hyperfinite set: an internal set and an initial sequence of \( ^*\mathbb{N} \). Let us consider this hyperfinite set as the sample space of a lottery. The probability measure for this hyperfinite

\[11\]With Loeb measures it is possible to transform the non-standard co-domain to a standard one (Cutland, 1983), but this still leaves us with a non-standard domain instead of \( P(\mathbb{N}) \).
lottery can be obtained by Transfer of the standard measure of a finite lottery, given in equation (2.1). Thus we obtain the following internal probability measure:

\[
*P_\alpha : *\mathcal{P}(\{1, \ldots, \alpha\}) \to ^*\left[0, 1\right]_Q^\#(*A)_{\alpha}
\]

Because this probability measure is obtained as the star-map of a standard probability measure on a finite sample space, which is FA, \( *P_\alpha \) is HFA by Transfer.

As such, \( *P_\alpha \) cannot be regarded as a satisfactory solution to the problem of de-numerably infinite lotteries. The order type of the event space will be vastly different from the order type of \( \mathbb{N} \). Thus, to use terminology introduced earlier, this probability function operates at best with a relabelling of the natural numbers. We have seen in our discussion of LABEL that probability functions on infinite event spaces simply cannot be taken to be invariant under relabelling. So the “internalization” of the problem of infinite lotteries in the non-standard universe does not solve the original problem. Instead, it is a solution to a different problem.

Note that although \( \mathbb{N} \) is a subset of \( \{1, \ldots, \alpha\} \), it is an external one (cf. paragraph 2.4.6), so it is not part of the domain of \( *P_\alpha \). Therefore we cannot use conditionalization on \( *P_\alpha \) to define a probability measure on the sample space \( \mathbb{N} \).

If not by conditionalization, what is the relation between the hyperfinite lottery on \( \{1, \ldots, \alpha\} \) and the hyperrational probability measure on \( \mathbb{N} \)? Any subset \( A \) of \( \mathbb{N} \) has a hyperextension \( *A \in *\mathcal{P}(*\mathbb{N}) \); the intersection of \( *A \) with \( \{1, \ldots, \alpha\} \) is an internal subset of \( \{1, \ldots, \alpha\} \) and thus can be assigned a probability value by the internal measure \( *P_\alpha \). This is precisely the inner working of \( P_{num} \): a three-step-process that can be read off from equation (2.19).\(^{12}\)

### 2.6.4 Non-standard probability and asymptotic density

All frequencies that are experimentally accessible concern finite sequences of trials only, and they can be expressed as rational numbers. Limiting frequencies are an infinite idealization of observable frequencies resulting in real-valued probabilities. Our approach is a different idealization that leads to hyperrational probabilities. We will show that its standard part is exactly equal to the limiting frequency. Therefore it is impossible to favour one solution over the other on mathematical or experimental grounds. So we are addressing a question of epistemology: the approaches encapsulate a different vision on probability and it is up to us to decide which one is most in concord with our intuitions.

#### 2.6.4.1 Co-domain \( \mathbb{R} \) versus \( *\mathbb{Q} \)

For a finite lottery, the probability measure takes values on \( \mathbb{Q} \). However, this co-domain is not closed: a limit of a rational sequence is not always defined in \( \mathbb{Q} \), but it is in \( \mathbb{R} \). In general, the sample space may be infinite; to this end the range is extended to \( \mathbb{R} \). In contemporary presentations, probability theory is usually introduced as a

\(^{12}\)Our construction of \( P_{num} \) can be seen as a specific instantiation of the general suggestion made in (Benci et al., 2010, section 5).
special case of measure theory (e.g. Dudley, 2004); in this context, the idea that a probability measure is real-valued—which originally was part of an axiom (axiom III in Kolmogorov, 1933, p. 2)—is so basic that it remains a tacit assumption.

Using NSA, the natural extension of the co-domain is \( \ast \mathbb{Q} \) rather than \( \mathbb{R} \). In order to assign sizes to all finite subsets of \( \mathbb{N} \), we already need all of \( \mathbb{N} \) (and zero), and to assign probabilities to all finite lotteries, we need all numbers in \([0,1]_{\mathbb{Q}}\). Thus, in order to assign sizes to all subsets of \( \mathbb{N} \), finite as well as infinite, a larger set is required, which can be obtained using NSA: we consider \( \ast \mathbb{N} \) (and zero), or at least an initial part of it, \( \{0,1,\ldots,\alpha\} \) with \( \alpha \in \ast \mathbb{N} \setminus \mathbb{N} \). To assign probabilities to a countably infinite lottery (or all hyperfinite ones), we need \([0,1]_{\ast \mathbb{Q}}\) as the range.

Although NSA can be introduced axiomatically, for instance using alpha-theory, and thus need not be intrinsically more difficult than learning standard analysis, the latter was developed first and is still much more common. There are at least two reasons why mathematicians generally prefer the standard framework of classical analysis over the framework of nonstandard analysis. Firstly, nonstandard analysis was rigorously developed much later than standard analysis. So, even though it is not intrinsically more difficult to learn nonstandard analysis, it has the disadvantage of unfamiliarity. Secondly, because of the Transfer Principle, non-standard analysis does not yield any new information about the standard real numbers. However, the concept of a fair infinite lottery simply begs for the co-domain of the sought-for probability functions to be modelled using standard numbers. Thus the Transfer Principle cannot be used here to transport us back to the familiar shore of the standard real numbers. But we do stress that the standard but ultimately unsatisfactory \( \mathbb{R} \)-solution can be interpreted as the standard part of any suitable \( \ast \mathbb{Q} \)-function.

### 2.6.4.2 Standard valued approximations and the failure of SUM

As we know from section 2.3, asymptotic density is only FA. Admittedly, CA is not achieved by the \( \ast \mathbb{Q} \) approach either, but the latter tells us why this is not so: the extension of FA for finite lotteries to the countably infinite case does not lead to CA, but rather to HFA (or HCA in a trivial sense). Or, put differently, this solutions shows us that we can extend our finite SUM intuition to HFA. We now investigate how this relates to the FA of asymptotic density.

Let us approximate the hyperreal-valued probability measure (up to infinitesimals) by a real-valued measure. To this end, NSA provides the standard part function, \( st \) (paragraph 2.4.4). First, we will illustrate this by three examples concerning the lottery on \( \mathbb{N} \):

- **Single ticket:** \( A = \{n\} \) for some \( n \in \mathbb{N} \). Then \( P_{\text{num}}(A) = \text{num}(\{n\})/\alpha = 1/\alpha \). Since \( \alpha \) is an infinite hypernatural, its inverse is an infinitesimal with standard part zero: \( st(P_{\text{num}}(A)) = 0 \).

- **All but one ticket.** \( B = \mathbb{N} \setminus \{n\} \). Then \( P_{\text{num}}(B) = \text{num}(B)/\alpha = (\alpha - 1)/\alpha = 1 - 1/\alpha \). So \( st(P_{\text{num}}(B)) = 1 \).

- **Consider an arbitrary \( m \in \mathbb{N} \). For all \( k \in \{0,1,\ldots,m-1\} \), let \( C_k \) be the set \( \{n \mid n \mod m = k\} \). Then \( st(P_{\text{num}}(C_k)) = st(1/m) = 1/m \) for all \( k \).**
In the examples, the standard part of the hyperrational-valued probability measure $P_{num}$ equals the real-valued asymptotic density. This close connection should not come as a surprise, since $P_{num}$ and $P_{ad}$ are, respectively, the ultrafilter equivalence class (or alpha-limit) and the Hahn-Banach limit of one and the same $\omega$-sequence. In fact, what we have observed in the examples holds in general:\textsuperscript{13}

\[ st \circ P_{num} = P_{ad} \quad \text{(2.30)} \]

**Proof.**

Fix an arbitrary $A \in \mathcal{P}(\mathbb{N})$. We need to show that $st(P_{num}(A)) = P_{ad}(A)$.

We know that $P_{ad}(A)$ is the Hahn-Banach limit of $(a_n) = \left\{ \frac{\#(A \cap \{1, \ldots, n\})}{n} \right\}$ (eq. 2.3). From equation (2.9), we know that in order to proof that $st(P_{num}(A))$ is the Hahn-Banach limit of this sequence, we need to find $H_0, H_1 \in ^*\mathbb{N}\setminus\mathbb{N}$ with $H_0 < H_1$ and an internal sequence of hyperreals $h_{H_0}, \ldots, h_{H_1}$ such that $\sum_{N=H_0}^{H_1} h_{N,N} = 1$ and:

\[ st(P_{num}(A)) = st \left( \sum_{N=H_0}^{H_1} h_{N,N}^* a_N \right) \]

First transform the left hand side using equations 2.6 and 2.21:

\[ P_{num}(A) = \left[ \frac{\#(A \cap \{1, \ldots, n\})}{n} \right]_U = a_\alpha \]

Then we see that the above equality is equivalent to:

\[ st(a_\alpha) = st \left( \sum_{N=H_0}^{H_1} h_{N,N}^* a_N \right) \]

which holds if we choose $H_0 < \alpha$, $H_1 \geq \alpha$, $h_{N,N} = 0$ for all $N \neq \alpha$ and $h_{\alpha} = 1$.\textsuperscript{14} \hfill \Box

This relation elucidates why (generalized) asymptotic density has poorer additivity properties than the non-standard measure: each time the standard part function is applied, a rounding-off error (up to infinitesimals) is introduced. When adding a finite number of rounded values this still only introduces an infinitesimal error, but when adding an infinite number of approximated terms, the total error may be appreciable. In particular, when summing over the family of all singletons, we see SUM fail most dramatically: the total error is as high as $\alpha \cdot 1/\alpha = 1$, which is 100% error in case of probabilities.

\textsuperscript{13}This is true for any $P_{num}$ in the family of solutions corresponding to the freedom of choice in the free ultrafilter.

\textsuperscript{14}For the special case where $(a_n)$ converges, $st(a_\alpha) = \lim_{n \to \infty} a_n$ (eqs. 2.6 and 2.8). Thus the proof for that case boils down to: $st(P_{num}(A)) = \lim_{n \to \infty} a_n = P_{ad}(A)$
2.7 Conclusion

In this chapter, we have argued that fair infinite lotteries can best be described using tools and concepts of non-standard analysis. We have constructed a uniform probability measure which is defined on the full power set algebra of \( \mathbb{N} \), and which takes its values in the non-standard extension of the \([0, 1]_{\mathbb{Q}}\)-interval. The construction is closely related to that of asymptotic density measures. The resulting probability measure is uniform, it gives a non-zero probability to the event that a given ticket wins, and it is not just finitely but also infinitely additive.

Lewis (1986a) argued that every possibility should be assigned a non-zero probability. This has been taken by him and others as a reason for advocating probability functions that take their values in a non-standard \([0, 1]\)-interval. We do not advocate the principle of Regularity in general. But we have seen that in the context of the lottery on the natural numbers, it follows from desiderata that we are committed to on independent grounds. So it was natural for us to try to work out a non-standard approach in some detail.

Proposals to construct non-standard measures on infinite sample spaces have been made in the literature (Cutland, 1983). But instead of taking the natural numbers as their domain, these probability functions operate on a non-standard extension of the natural numbers. This entails that they do not provide a genuine solution to the problem of lotteries on the natural numbers; instead, they change the problem. The probability measure that we have constructed is of mixed origins. It stands with one leg in the classical universe, and with its other leg in the non-standard universe.

Asymptotic density and our probability measure are based on the same sequence of partial fractions, and both look at its behavior ‘at infinity’. Only the formalization of this statement is achieved differently: by (a generalization of) classical limits (Schurz and Leitgeb, 2008) in the first case, and by free ultrafilter-based equivalence classes in the second. Asymptotic density takes values in the standard \([0, 1]\)-interval. The price for this is that such probability functions can only be finitely additive for fair lotteries on \( \mathbb{N} \).

The solution that we propose meets the conceptual and intuitive requirements connected with a lottery on \( \mathbb{N} \): it is fair, uniform, defined on the whole power set algebra of \( \mathbb{N} \), and infinitely additive. We have seen that we do not end up with a unique solution. The probability functions that we propose are only determined up to the choice of an ultrafilter. As far as we can presently see, no choice of ultrafilter is superior to any other. We do not exclude that when more intuitive constraints are imposed, the class of satisfactory probability functions can be narrowed down further. But at present we see no way of doing so.

The most notable feature of our solution is that whereas it is infinitely additive, its additivity behavior is not adequately described by summing over the natural numbers, but by summing over a non-standard extension of the natural numbers. One might naively expect the relevant sum generally to be an \( \omega \)-sum of non-standard weights. But it emerged that the probability of an \( \omega \)-length family of events will be a sum of a much longer order type. In this context, it should be recalled that even Kolmogorov, the discoverer of the classical \( \sigma \)-additivity axiom, emphasized that the
condition is merely a useful assumption of idealization but is not contained in the meaning of probability.\footnote{Kolmogorov (1933, p. 15) states: “Infinite fields of probability occur only as idealized models of real random processes. \emph{We limit ourselves, arbitrarily, to only those models which satisfy Axiom VI.}” (Emphasis in the original.) In Kolmogorov’s paper, ‘field’ means ‘algebra’ and ‘Axiom VI’ refers to the ‘Axiom of Continuity’ which, together with FA, automatically leads to CA in cases where the algebra is a $\sigma$-algebra.} It turned out that in the context of lotteries on the natural numbers, $\sigma$-additivity is not the right idealization.

The obvious next step would be to consider larger, nondenumerable lotteries, such as lotteries on $\mathbb{R}$. The description of fair lotteries on $\mathbb{R}$ is a mathematically significantly more difficult affair, and will be left for future work.