Minimizing the Cost of Guessing Games

Lindsay Czap
Grand Valley State University

Follow this and additional works at: http://scholarworks.gvsu.edu/honorsprojects

Part of the Physical Sciences and Mathematics Commons

Recommended Citation
Czap, Lindsay, "Minimizing the Cost of Guessing Games" (2016). Honors Projects. 572.
http://scholarworks.gvsu.edu/honorsprojects/572
Minimizing the Cost of Guessing Games

David Clark, Lindsay Czap

Department of Mathematics, Grand Valley State University, Allendale, MI

Students: czapl@mail.gvsu.edu, lczap@bgsu.edu
Mentor: clarkdav@gvsu.edu

ABSTRACT
A two-player “guessing game” is a game in which the first participant, the “Responder,” picks a number from a certain range. Then, the second participant, the “Questioner,” asks only yes-or-no questions in order to guess the number.

In this paper, we study guessing games with lies and costs. In particular, the Responder is allowed to lie in one answer, and the Questioner is charged a cost based on the content of each question.

Guessing games with lies are closely linked to error correcting codes, which are mathematical objects that allow us to detect an error in received information and correct these errors. We will give basic definitions in coding theory and show how error correcting codes allow us to still guess the correct number even if one lie is involved.

We will additionally seek to minimize the total cost of our games. We will provide explicit constructions, for any cost function, for games with the minimum possible cost and an unlimited number of questions. We also find minimum cost games for games with a restricted number of questions and a constant cost function.

KEYWORDS
Ulam’s Game; Guessing Games With Lies; Error Correcting Codes; Pairwise Balanced Designs; Steiner Triple Systems

INTRODUCTION
The game “20 Questions” has always been a staple for children trapped in long car rides. This game involves two players: the first player picks an object and the second player asks “yes or no” questions in order to guess which object the first picked. The game alternates between questions and answers so questions may be constructed based off of previous answers. This type of adaptive guessing game is called an online game. For example, you would not ask the question “Does it have four legs?” if you have already received a “yes” to the question “Is your object a vegetable?”

“20 Questions” requires both players to be present and for each response to be given immediately after its corresponding question. In this paper, we will focus on a different type of guessing game: In an offline guessing game, the questioner must ask all of the questions at once and then receive all of the responses at once afterwards. This means that we must construct all of the questions ahead of time and we cannot adapt questions based upon previous responses. These types of non-adaptive guessing games are called offline games and are the focus of this paper.

In our offline guessing games, the two players decide on a range of m integers starting at 1. The first player, the “Responder,” secretly chooses one of those integers x. The second player, the “Questioner,” then constructs yes-or-no questions of the form “Is x ∈ S?” where S ⊆ [1, m]. Since these games are offline, all of the questions that the Questioner wishes to ask the Responder will be asked at once and then all of the answers will be received at once.

Example 1. The following is an example of an offline guessing game with m = 8 possible answers.

Questioner:

Question 1: “Is your number in {5, 6, 7, 8}?”
Question 2: “Is your number in {3, 4, 7, 8}?”
Question 3: “Is your number in {2, 4, 6, 8}?”

Responder:
The Questioner can then use the Responder’s responses to figure out which integer was chosen. A response of “yes” to Question 1 eliminates the possibilities $\{1, 2, 3, 4\}$, leaving only $\{5, 6, 7, 8\}$ as possibilities. A response of “yes” to Question 2 eliminates the possibilities $\{5, 6\}$, leaving only $\{7, 8\}$. A response of “no” to Question 3 eliminates the possibility $\{8\}$. Thus, the Questioner knows that the Responder originally chose 7 as their integer.

The guessing games that we will be analyzing have a special property: The Responder is allowed to lie to the Questioner. However, even though the Responder is allowed to lie, these games are designed so that the Questioner can still guess the correct number that the Responder has chosen. We will also add the idea of cost to a guessing game: The Questioner will be charged a certain cost whenever they ask a question. In the following sections, we will give precise definitions and background for each of these additional twists.

**BACKGROUND**

In this section, we give background and definitions, leading up to a precise definition of guessing games with lies and cost functions.

Guessing games with lies have been extensively studied. The idea of a “searching game” with a lie was suggested by Ulam in his autobiography \(^1\), which in turn spawned a large and active field of research. In general, work in this area has focused on minimizing the number of questions needed to determine the Responder’s number correctly. In particular, Spencer \(^2\) focused on solving “Ulam’s game” with 1,000,000 numbers and 1 lie, which turns out to be the equivalent of the “20 questions” game with 1 lie permitted. Offline guessing games with lies turn out to be equivalent to important objects in mathematics called *error-correcting codes*. See \(^3\^4\) for some excellent surveys of the literature on guessing games with lies.

Our particular focus in this paper will be adding the idea of a “cost function” to offline guessing games with lies. A guessing game $G$ with cost function $c$ is an offline guessing game in which each of the possible answers $i$ is assigned a “cost” $c(i)$. For each question, the Questioner will be “charged” the cost of all of the answers that they are asking about. For example, the cost of the question “Is your number in $\{1, 3\}$?” will be the sum of the costs of 1 and 3. This generalizes the idea of minimizing the number of questions required, and allows us to measure the “efficiency” of a game in a more general way.

**Representations of guessing games**

For our offline guessing games, we will represent the answers to each of our questions using binary vectors. A binary vector is an ordered list of 1’s and 0’s. We will assign the digit 0 to represent an answer of “no”, and the digit 1 will represent “yes”. For example, if we asked 5 consecutive questions and their respective answers were “yes”, “yes”, “no”, “yes”, “no”, we would represent these responses with the binary vector $(1, 1, 0, 1, 0)$. We will often use the shorthand notation: $11010$.

We can think of an answer vector as the “fingerprint” of the Responder’s secret number. It contains all of the information necessary to determine the secret number. In particular, in every guessing game, every secret number must have a different answer vector or else it would be impossible to distinguish between some of the numbers.

**Definition 2.** For a given natural number $m$, a *cost function* $c$ is a function $c : \{1, 2, \ldots, m\} \to \mathbb{R}^+$ with $c(1) \leq c(2) \leq \cdots \leq c(m)$. For simplicity, we often represent $c$ as a vector: $c = (c(1), c(2), \ldots, c(m))$.

We can now define the fundamental object in this paper:

**Definition 3.** An $(m, c, \ell)$ *offline guessing game $G$ with $n$ questions* is an ordered list of $n$ questions of the form “Is your number in the set $S_i$?”, where $S_i \subseteq \{1, 2, \ldots, m\}$, together with a cost function $c = (c(1), c(2), \ldots, c(m))$. The questions are arranged so that the Questioner can determine the Responder’s secret number, even if the Responder is allowed to lie in at most $\ell$ answers.

We will often omit the word “offline”, since we will only be working with offline guessing games in this paper. Note that we have not yet indicated *how* the questions may be arranged to let the Questioner determine the Responder’s secret number.
in the face of lies. We will address this in the following sections.

Next, we give a more precise definition of the “answer vectors” described at the beginning of this section.

**Definition 4.** Given an \((m, c, \ell)\) guessing game \(G\) with \(n\) questions, the **answer vector** for \(i\), denoted \(v_i\), is the binary vector \(v_i \in \mathbb{F}_2^n\) consisting of the Responder’s truthful answers to the \(n\) questions, in order, assuming that the Responder’s secret number is \(i\). The **answer vector set of \(G\)**, denoted \(V_G\), is the set \(\{v_1, v_2, \ldots, v_m\} \subseteq \mathbb{F}_2^n\) of answer vectors for \(i = 1, 2, \ldots, m\).

**Example 5.** Consider the set \(V_G = \{00000, 01110, 10101, 11011\} \subseteq \mathbb{F}_2^5\). This is a set of 4 answer vectors of length 5 for the numbers \(\{1, 2, 3, 4\}\). This could be the answer vector set for a \((4, c, 1)\) guessing game with 5 questions. In particular, we have \(v_1 = 00000, v_2 = 01110, v_3 = 10101\), and \(v_4 = 11011\).

**Definition 6.** Given an \((m, c, \ell)\) game \(G\) with \(n\) questions, the **guessing game matrix** of \(G\) is an \(m \times n\) binary matrix denoted \([G]\) in which row \(j\) corresponds to vector \(v_j \in V_G\). That is, the \(j\)th row of \([G]\) is the list of the Responder’s truthful answers to the \(n\) questions in order, assuming that the Responder’s secret number is \(j\).

Answer vectors and guessing game matrices are the main ways in which we will study guessing games in this paper. Because the rows of a guessing game matrix are the same as the answer vectors, we will often refer to the rows of a matrix as “vectors.”

**Example 7.** There are many different vector sets that could be used to construct a \((4, c, 1)\) game. **Figure 1** shows one possible game matrix \([M]\) for a \((4, c, 1)\) game \(M\) with 5 questions. We created this matrix by choosing \(V_M = \{00000, 01110, 10101, 11011\}\).

\[
[M] = \begin{bmatrix}
  0 & 0 & 0 & 0 & 0 \\
  2 & 0 & 1 & 1 & 1 \\
  3 & 1 & 0 & 1 & 0 \\
  4 & 1 & 1 & 0 & 1 \\
\end{bmatrix}
\]

**Figure 1.** A possible game matrix, \([M]\), for a \((4, c, 1)\) game.

Similarly, **Figure 2** shows another possible matrix \([N]\) for a \((4, c, 1)\) game \(N\) with 7 questions. We created this second matrix by choosing \(V_N = \{1110110, 0101011, 1011110, 0101100\}\).

\[
[N] = \begin{bmatrix}
  1 & 1 & 1 & 1 & 0 & 1 & 0 \\
  2 & 0 & 1 & 0 & 1 & 0 & 1 \\
  3 & 1 & 0 & 1 & 1 & 1 & 0 \\
  4 & 0 & 1 & 0 & 1 & 0 & 0 \\
\end{bmatrix}
\]

**Figure 2.** Another possible game matrix, \([N]\), for a \((4, c, 1)\) game.

Both of the guessing game matrices \([M]\) and \([N]\) are created by choosing valid sets of representative binary vectors and therefore are both valid matrices representing a \((4, c, 1)\) game.

It is possible to determine a game’s questions entirely from its guessing game matrix. In **Figure 3**, we have highlighted the 4th column of matrix \([M]\) from the previous example. We say that this highlighted column is the question vector \(q_4 = (0, 1, 0, 1)\). We know that, if responding truthfully, the Responder will answer “yes” to the 4th question if and only if there is a 1 in the row of \([M]\) corresponding to their secret number. Therefore, \(q_4\) would be the 4th question in the game \(G\) and would be asked as: “Is your number in \(\{2, 4\}\?” because these are the answers whose answer vector have a 1 in column 4 of \([M]\).

For this reason, we will often treat a guessing game not as a list of questions, but rather as a list of answer vectors. These two viewpoints are exactly equivalent.
Cost functions

We will now focus on the effect of the cost function in a guessing game. The cost function will be our measure of the “efficiency” of a game.

**Definition 8.** Let $G$ be an $(m, c, \ell)$ guessing game. Suppose that $Q = \{p_1, p_2, \ldots, p_k\} \subseteq \{1, 2, \ldots, m\}$ is a question in $G$. Then the cost of question $Q$ is defined to be the sum of the costs of the elements of $Q$, that is,

$$c(Q) = \sum_{i=1}^{k} c(p_i).$$  \hspace{1cm} \text{Equation 1.}

Furthermore, the total cost of $G$, denoted $k_G$, is the sum of the cost of $Q$ for all questions $Q$ in $G$, that is,

$$k_G = \sum_{\text{question } Q} c(Q).$$  \hspace{1cm} \text{Equation 2.}

Because our guessing games are offline, all of the questions will always be asked. Thus the total cost of a game is a reasonable measure of the overall “efficiency” of a game. The following lemma gives a convenient way to calculate the total cost of a game in terms of its answer vectors.

**Lemma 9.** The total cost $k_G$ of an $(m, c, \ell)$ guessing game $G$ with answer vector set $\{v_1, v_2, \ldots, v_m\}$ is $k_G = c(1)|v_1| + c(2)|v_2| + \ldots + c(m)|v_m|$.

**Proof.** An answer vector $v_i$ has a 1 in coordinate $j$ if and only if the $j$th question asks about $i$. Thus number $i$ is asked about once for every 1 in each $v_i$, at a cost of $c(i)$ in each question. \hfill \Box

**Example 10.** Figure 4 shows a game matrix $[G]$ that defines a $(4, c, 1)$ game $G$ with $c = (2, 4, 6, 8)$.

![Figure 4](image_url)
Using Lemma 9, the total cost of $G$ is calculated by:

$$k_G = |v_1|c(1) + |v_2|c(2) + |v_3|c(3) + |v_4|c(4)$$

$$= 3c(1) + 3c(2) + 2c(3) + 2c(4)$$

$$= 3 \cdot 2 + 3 \cdot 4 + 2 \cdot 6 + 2 \cdot 8$$

$$= 46.$$ 

This demonstrates the essence of Lemma 9: It is possible to count the costs of a game by using the columns of its guessing game matrix (questions), or by using its rows (answer vectors).

**Error-correcting codes and lies**

In order to discuss the ability for our $(m, c, \ell)$ guessing games to detect and correct for lying, we must first introduce some ideas from a branch of mathematics known as coding theory. Let $\mathbb{F}_2^n$ be the space of all vectors of length $n$ over the finite field $\mathbb{F}_2$. We refer the interested reader to Huffman and Pless\(^5\) for further definitions and examples.

**Definition 11.** The weight of a vector $v \in \mathbb{F}_2^n$ is the number of nonzero coordinates in $v$.

**Definition 12.** The Hamming Distance between two vectors $x, y \in \mathbb{F}_2^n$, denoted $d(x, y)$, is the number of coordinates in which $x$ and $y$ differ.

**Example 13.** Suppose two integers $i, j$ have answer vectors $v_i = 1101$ and $v_j = 1011$. These vectors differ in 2 positions as shown in Figure 5. Thus $d(v_i, v_j) = 2$.

<p>| | | | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td>$i$</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$j$</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>1</td>
</tr>
</tbody>
</table>

**Figure 5.** Answer vectors $v_i$ and $v_j$.

**Definition 14.** The minimum distance $d(C)$ of a set of vectors $C \subseteq \mathbb{F}_2^n$ is the smallest Hamming Distance between any two distinct vectors in $C$.

With these ideas in hand, we are ready to define the fundamental idea that will allow us to detect and correct lies in guessing games.

**Definition 15.** An $(n, M, d)$ error-correcting code $C$ (or “ECC”) is a subset of $\mathbb{F}_2^n$ of size $M$ with $d(C) = d$. We call these vectors in $C$ codewords. An $(n, M, d)$ error-correcting code can be represented by an $M \times n$ matrix denoted $[C]$ whose rows are the codewords of $C$.

An ECC has the capacity to detect and then correct for errors that can occur in a codeword when it is exposed to contextual factors. These factors can include noise over a digital channel, a misprint in written text, or – in the case of our guessing games – a player lying. The capacity for an ECC to detect and correct for errors is based on the minimum distance of that code.

The process by which minimum distance allows a code to correct errors is called nearest-neighbor decoding. For an $(n, M, d)$ code $C$, a message vector $m$ is any one codeword. It is transmitted through a channel, where errors occur in the form of bits “flipped” from 1 to 0 or vice versa. The received message vector $r$ is then a binary vector with $n$ coordinates that may share some entries with $m$. To decode this message, vector $r$ is compared to all of the codewords in the code. The codeword $c$ that minimizes $d(c, r)$ is chosen as the “nearest neighbor” to $r$. In particular, if there are at most $\ell$ mistakes in $r$, and the minimum distance of a code is at least $2\ell + 1$, then there must be a unique codeword that is “closest” to $r$ and which results in a correct decoding.

**Proposition 16** (See Huffman and Pless\(^5\)). An $(n, M, d)$ error-correcting code $C$ with $d \geq 2\ell + 1$ can detect and correct for up to $\ell$ errors.
Example 17. Figure 6 shows the vectors in an error-correcting code $C$, written as the rows of $C$’s matrix. It follows that $d(C) = 3$ since the Hamming Distance between any two codewords in $C$ is at least 3. Thus, $C$ can detect and correct for up to $\ell = 1$ errors.

For example, if the vector $\mathbf{r} = (1, 1, 0, 1, 0, 0)$ were received, we could compare its distance to each of the four vectors in $C$. The 4th vector has a distance of 1 to $\mathbf{r}$, and no other vector has a distance that is 1 or lower, and so $\mathbf{r}$ would be corrected as $(1, 1, 0, 0, 0, 0)$.

\[
\begin{array}{cccccc}
1 & 0 & 0 & 1 & 1 & 0 \\
0 & 1 & 1 & 1 & 0 & 1 \\
1 & 0 & 1 & 0 & 0 & 1 \\
1 & 1 & 0 & 0 & 0 & 0 \\
\end{array}
\]

Figure 6. The code $C$.

A valid guessing game matrix $[G]$ for an $(m, c, \ell)$ game $G$ with $n$ questions can be viewed as the matrix for an $(n, M, 2\ell + 1)$ error-correcting code and vice versa. The binary answer vectors in $[G]$ that are assigned to each of the $m$ possible integers correspond to the codewords of the code represented by $[G]$. The number of lies $\ell$ that are allowed in $G$ corresponds to the number of errors that the corresponding ECC can detect and correct. By this, it follows that in order for $[G]$ to be a valid guessing game matrix for an $(m, c, \ell)$ game $G$, it must be true that $d([G]) \geq 2\ell + 1$. Thus, trying to find an $(m, c, \ell)$ offline guessing game $G$ with $n$ questions is equivalent to finding an $(n, m, 2\ell + 1)$ error-correcting code.

In the remainder of this paper, we will find $(m, c, 1)$ games with $n$ questions by solving the equivalent problem of finding $(n, m, 3)$ error-correcting codes. However, we will have the additional requirement of minimizing the total cost of the corresponding game, a restriction which is not normally considered when constructing an error-correcting code.

RESULTS

Guessing Games with an Unrestricted Number of Questions

In this section, we will make several assumptions. First, we restrict our investigations to guessing games allowing for only one lie. Second, we place no restriction on the number of questions that the Questioner asks. Thus, for each guessing game we will specify only the number of possible answers $m$, and will leave the number of questions $n$ unrestricted. Instead, we will rely on the cost function for our games to measure the efficiency of the game.

For any $(m, c, 1)$ guessing game $G$, either the all-zeroes vector $\mathbf{0}$ is in the set of answer vectors $Y_G$, or it isn’t. We will show two guessing game matrix constructions that we can use to minimize the total cost of a guessing game, depending upon the inclusion or exclusion of $\mathbf{0}$.

Definition 18. An $m \times n$ $(3, 0)$ matrix $[G]_{m}^{(3,0)}$ is an $m \times (2m + 1)$ binary matrix as shown in Figure 7.

\[
\begin{array}{cccccccc}
1 & 1 & 1 & 0 & 0 & 0 & \ldots & 0 \\
2 & 0 & 0 & 1 & 1 & 1 & 0 & \ldots & 0 \\
\vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\
m-1 & 0 & 0 & 0 & \ldots & 0 & 1 & 1 & 1 \\
m & 0 & 0 & 0 & \ldots & 0 & 0 & 0 & 0 \\
\end{array}
\]

Figure 7. An $m \times n$ $(3, 0)$ matrix $[G]_{m}^{(3,0)}$.

That is, there are $m - 1$ rows of weight 3 and a single row of weight 0 at the bottom. The rows of weight 3 are arranged in a “staircase” fashion so that each row shares a 1 in exactly one coordinate with the previous row, and in exactly one coordinate with the subsequent row.
Definition 19. An $m \times n$ $(2, 1)$ matrix $[G]_{m}^{(2,1)}$ is an $m \times (2m-1)$ binary matrix as shown in Figure 8.

That is, there are $m-1$ rows of weight 2 and a single row of weight 1 at the bottom. The rows of weight 2 do not share any nonzero coordinates, and the row of weight 1 shares its single 1 with no other row.

Note that, by construction, the rows of both $(3,0)$ and $(2,1)$ matrices have a minimum distance of 3.

Lemma 20. Let $m$ and $c = (c(1), c(2), \ldots, c(m))$ be given. Among all $(m, c, 1)$ guessing games that contain an answer vector consisting of all 0’s, the minimum total cost is $k_{0 \in V} = 3c(1) + 3c(2) + \ldots + 3c(m-1)$. A game with guessing game matrix $[G]_{m}^{(3,0)}$ will attain this minimum total cost.

Proof. Let $m$ and $c = (c(1), c(2), \ldots, c(m))$ be given. Let $[G]$ be the guessing game matrix of the $(m, c, 1)$ game that attains the minimum total cost among all $(m, c, 1)$ games that contain a zero answer vector. By assumption, $[G]$ must contain a vector of weight 0. Because $d([G]) \geq 3$, all other rows of $[G]$ must have weight at least 3. Using Lemma 9, the lowest possible cost occurs if all $m-1$ nonzero rows of $[G]$ are of weight 3.

Next we claim that the row of $[G]$ corresponding to answer $m$ must be 0. (Recall that, by assumption, $c(m)$ is the largest cost.) To prove this, we let $\{v_1, v_2, \ldots, v_m\}$ be a fixed set of vectors of weight 3 and minimum distance at least 3. Suppose we let $k_i$ be the total cost of a game with question vectors when $v_i, i \neq m$, is replaced with 0 and $k_m$ be the total cost when $v_m$ is replaced with 0. It then follows that

$$k_i = 3c(1) + 3c(2) + \ldots + 3c(i-1) + 3c(i+1) + \ldots + 3c(m)$$

and

$$k_m = 3c(1) + 3c(2) + \ldots + 3c(m-2) + 3c(m-1).$$

Thus,

$$k_i - k_m = 3c(m) - 3c(i).$$

Since $c(i) \leq c(m)$ by assumption, it then follows that $k_i - k_m \geq 0$ and therefore $k_i \geq k_m$. Thus it must be that $[G]$ contains a single vector of weight zero that represents the integer $m$, and all other vectors have weight 3.

The $(3,0)$ matrix $[G]_{m}^{(3,0)}$ meets these criteria exactly, and so it must represent the minimum possible cost. The cost of this matrix is exactly $k_{0 \in V} = 3c(1) + 3c(2) + \cdots + 3c(m-1)$. ☐

Lemma 21. Let $m$ and $c = (c(1), c(2), \ldots, c(m))$ be given. Among all $(m, c, 1)$ guessing games that do not contain an answer vector consisting of all 0’s, the minimum total cost is $k_{0 \in V} = 2c(1) + 2c(2) + \ldots + 2c(m-1) + c(m)$. A game with guessing game matrix $[G]_{m}^{(2,1)}$ will attain this minimum total cost.

Proof. Let $m$ and $c = (c(1), c(2), \ldots, c(m))$ be given. Let $[G]$ be the guessing game matrix of the $(m, c, 1)$ game that attains the minimum total cost among all $(m, c, 1)$ games that do not contain a zero answer vector.

We first claim that $[G]$ will contain a vector of weight 1. Suppose such a vector does not exist in $[G]$. It would follow that every vector in $[G]$ has a weight of at least 2. We will demonstrate that some vector may be reduced to a weight of 1. Suppose the set of vectors $\{v_1, \ldots, v_i, \ldots, v_m\}$ in Figure 9a represent the rows of $[G]$. These rows satisfy $d([G]) \geq 3$ and all
have weight of at least 2. Replace \( v_i \) with \( 0 \) and ask an additional question about only \( i \), as illustrated in Figure 9b. We can denote this new answer vector by \( 0\bar{v}_j \). It follows that \( d(0, v_j) \geq 2 \) for all \( v_j \in V_G \), and because the single additional 1 occurs in a column of all 0’s, \( d(0\bar{v}_j, v_j) \geq 3 \). Thus the original \( v_i \) can be changed to \( 0\bar{v}_j \) while maintaining \( d([G]) \geq 3 \).

<table>
<thead>
<tr>
<th>( v_1 )</th>
<th>( v_i )</th>
<th>( v_m )</th>
</tr>
</thead>
<tbody>
<tr>
<td>...</td>
<td>...</td>
<td>...</td>
</tr>
</tbody>
</table>

(a) The rows of \([G]\) that satisfy \( d([G]) \geq 3 \).

Next we claim that \([G]\) contains exactly one vector of weight 1. Suppose two vectors of weight 1 exist. It would then follow that their Hamming Distance would be at most 2 and thus \( d([G]) \leq 2 \). Therefore \([G]\) contains exactly one vector of weight 1.

Since \([G]\) contains one vector of weight 1, to ensure that \( d([G]) \geq 3 \), it follows that all other vectors must have a weight of at least 2. Using Lemma 9, the lowest possible cost then occurs if all \( m - 1 \) rows of weight at least 2 have weight exactly 2. Thus \([G]\) will contain exactly one vector of weight 1 and \( m - 1 \) vectors of weight 2.

The \((2, 1)\) matrix \([G]_{(2,1)}\) meets these criteria exactly, and so it must represent the minimum possible cost. The cost of this matrix is exactly \( k_{0\not\in V} = 2c(1) + 2c(2) + \ldots + 2c(m - 1) + c(m) \).

We are now ready to describe the minimum-cost guessing games, where the number of questions is unrestricted.

**Theorem 22.** Let \( m \) and \( c = (c(1), c(2), \ldots, c(m)) \) be given.

- If \( c(m) < \sum_{i=1}^{m-1} c(i) \) then the minimum cost of a \((m, c, 1)\) guessing game is realized by a \((2, 1)\) matrix \([G]_{(2,1)}\).
- If \( c(m) > \sum_{i=1}^{m-1} c(i) \) then the minimum cost of a \((m, c, 1)\) guessing game is realized by a \((3, 0)\) matrix \([G]_{(3,0)}\).
- If \( c(m) = \sum_{i=1}^{m-1} c(i) \) then the costs of a game realized by a \((3, 0)\) matrix and a \((2, 1)\) matrix are both equal to the minimum cost.

**Proof.** Let \( m \) and \( c = (c(1), c(2), \ldots, c(m)) \) given. Let \( k_{(3,0)} \) and \( k_{(2,1)} \) be the total cost of an \((m, c, 1)\) guessing game realized by a \((3, 0)\) matrix and a \((2, 1)\) matrix respectively.

By Lemmas 20 and 21, either a \((3, 0)\) or a \((2, 1)\) matrix gives the minimum total cost of an \((m, c, 1)\) game, depending on whether or not a zero vector is present. Thus one of these two must represent the minimum possible cost of an \((m, c, 1)\) game.

Consider the case where \( c(m) < \sum_{i=1}^{m-1} c(i) \). Using our assumptions and calculating the total cost of a \((2, 1)\) and \((3, 0)\) matrix, we have:

\[
k_{(2,1)} = 2c(1) + 2c(2) + \cdots + c(m-1) + c(m) \\
< 2c(1) + 2c(2) + \cdots + c(m-1) + c(m) + \sum_{i=1}^{m-1} c(i) \\
= 3c(1) + 3c(2) + \ldots + 3c(m-1) + 0c(m) \\
= k_{(3,0)}.
\]
Thus, \( c(m) < \sum_{i=1}^{m-1} c(i) \) implies \( k_{(2,1)} < k_{(3,0)} \). Therefore, a \((2, 1)\) matrix gives the cheapest possible total cost.

Similarly, \( c(m) > \sum_{i=1}^{m-1} c(i) \) implies \( k_{(2,1)} > k_{(3,0)} \), which case a \((3, 0)\) matrix gives the cheapest possible total cost.

If \( c(m) = \sum_{i=1}^{m-1} c(i) \) then these two matrices give equal total costs.

**Example 23.** Let \( m = 4 \). The two candidates for the cheapest \((4, c, 1)\) guessing game are given in Figure 10.

If \( c = (1, 2, 3, 4) \), then \( c(4) = 4 < 1 + 2 + 3 = \sum_{i=1}^{3} c(i) \). Theorem 22 predicts that a \((2, 1)\) matrix will be cheapest. We calculate: \( c([G_1]) = 3(1) + 3(2) + 3(3) = 18 \) and \( c([G_2]) = 2(1) + 2(2) + 2(3) + 1(4) = 16 \). Thus the \((2, 1)\) matrix does indeed realize the cheapest game.

If instead \( c = (1, 2, 3, 10) \), then \( c(4) = 10 > 1 + 2 + 3 = \sum_{i=1}^{3} c(i) \). Theorem 22 predicts that a \((3, 0)\) matrix will be cheapest. We calculate: \( c([G_1]) = 3(1) + 3(2) + 3(3) = 18 \) and \( c([G_2]) = 2(1) + 2(2) + 2(3) + 1(10) = 22 \). Thus the \((3, 0)\) matrix is cheaper in this case.

**Guessing Games with a Restricted Number of Questions**

In the previous section, we saw that the cheapest guessing games required us to create a very large number of questions. We will now look at guessing games where the number of questions are restricted. As a simplification, we will only consider the cost function \( c = (1, 1, \ldots, 1) \). We begin, as we did in the previous section, by defining a special type of matrix that will be used throughout our study of this topic.

**Definition 24.** An \( m \times n \) \((1, 2, 3)\) matrix \([G]\) has the form shown in Figure 11.

\[
\begin{bmatrix}
A & 0 \\
B & 0 \\
\end{bmatrix}
\]

Here \([A]\) is a matrix with rows of weight 3, and \([B]\) is a matrix with \( \lfloor \frac{n-1}{2} \rfloor \) rows of weight 2. Further, we require that the minimum distance of the entire matrix \([G]\) be at least 3.

Note that, to guarantee a minimum distance of at least 3, no row of weight 2 or 3 can have a 1 in the rightmost column. In addition, \( \lfloor \frac{n-1}{2} \rfloor \) is the largest number of rows of weight 2 that can be contained in a matrix with \( n - 1 \) columns and minimum distance at least 3. No two rows with weight 2 can share a 1 in any position.

In the following work, we will prove that \((1, 2, 3)\) matrices exist in certain cases. To do so requires some concepts from a branch of math known as Design Theory. The interested reader is referred to Lindner\(^6\) for more information beyond that.
presented below.

**Definition 25.** Let $X = \{1, 2, \ldots, u\}$ for $u \geq 3$ (these elements are called *varieties*), together with a set $B$ of subsets of $X$ of size 3 (called *blocks* or *triples*), such that every pair of varieties in $X$ occurs in exactly one block of $B$. The pair $(X, B)$ is a Steiner Triple System of $u$ varieties denoted $STS(u)$.

**Proposition 26.** An $STS(u)$ exists if and only if $u \equiv 1 \pmod{6}$ or $u \equiv 3 \pmod{6}$.6

**Definition 27.** The replication number $r$ of an $STS(u)$ is the number of blocks that contain a fixed variety.

It can be shown that for an $STS(u)$, that the replication number $r$ satisfies $r = \frac{u-1}{2}$, regardless of the variety chosen. Similarly, the number of blocks is exactly $b = \frac{u(u-1)}{6}$.

Steiner Triple Systems will provide the basis for constructing $(1, 2, 3)$ matrices. However, they exist only for a limited number of varieties. Thus we consider an alternative structure to cover some of the remaining cases.

**Definition 28.** Let $X = \{1, 2, \ldots, u\}$ for $u \geq 3$ (these elements are called *varieties*). Let $K$ be a non-empty set of natural numbers. Let $B$ be a set of non-empty subsets of $X$ called *blocks*, such that for each $b \in B$, $|b| \in K$. In addition, each pair of the $u$ elements of $X$ must appear in exactly one block of $B$. The pair $(X, B)$ is a $(u, K, 1)$ Pairwise Balanced Design, also denoted by $(u, K, 1)$-PBD.

Note that the key difference between an $STS(u)$ and a $PBD(u, K, 1)$ is that the sizes of the blocks in a $PBD(u, K, 1)$ may vary, while an $STS(u)$ always has blocks of size 3.

**Definition 29.** Let $D = (P, B)$ where $P = \{p_1, p_2, p_3, \ldots, p_u\}$ is a set of $u$ varieties and $B = \{b_1, b_2, b_3, \ldots, b_v\}$ is a set of subsets of $P$. The incidence matrix of $D$ is a $|B| \times |P|$ binary matrix $M$ in which each entry $M_{ij}$ contains a 1 if subset $b_i$ contains variety $p_j$ and a 0 if subset $b_i$ does not contain variety $p_j$.

**Example 30.** Figure 12 shows the incidence matrix of an $STS(7)$ with

$$B = \{\{1, 2, 4\}, \{2, 3, 5\}, \{3, 4, 6\}, \{4, 5, 7\}, \{5, 6, 1\}, \{6, 7, 2\}, \{7, 1, 3\}\}.$$

<table>
<thead>
<tr>
<th></th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
</tr>
</thead>
<tbody>
<tr>
<td>$b_1$</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_2$</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_3$</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>$b_4$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>$b_5$</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>$b_6$</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>$b_7$</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
</tr>
</tbody>
</table>

**Figure 12.** The incidence matrix of an $STS(7)$.

Note that all parameters of the $STS(u)$ are visible in this matrix. Each row represents one block of size 3, and so each row (that is, a block) has exactly three 1’s in it. Every pair of varieties appears in exactly one block, and so every pair of columns (that is, varieties) share a 1 in exactly one position. Finally, every variety is in exactly 3 blocks, and so every column contains exactly three 1’s as well.

**Example 31.** Figure 13 shows the incidence matrix of an $(11, \{3, 5\}, 1)$-PBD.

These incidence matrices will be the starting point from which we create $(1, 2, 3)$ matrices that will ultimately represent our guessing games.

**Lemma 32.** Let $n \equiv 1 \pmod{6}$ or $n \equiv 3 \pmod{6}$ and $2 \leq m \leq \frac{n(n-1)}{6} + 1$. Suppose $c = (1, 1, \ldots, 1)$. Then there exists an $m \times n$ $(1, 2, 3)$ matrix $[G]$ which is a valid guessing game matrix for an $(m, c, 1)$ guessing game $G$ with exactly $n$ questions.

**Proof.** Let $n \equiv 1 \pmod{6}$ or $n \equiv 3 \pmod{6}$. By Proposition 26, there exists a $STS(n)$ $S$ with varieties $N = \{1, 2, 3, \ldots, n-1, n\}$, block set $B$ such that $|B| = \frac{n(n-1)}{6}$, and replication number $r$ such that $r = \frac{n-1}{2}$. We will use $B$ to construct a new
Figure 13. The incidence matrix of an \((11, \{3, 5\}, 1)\)-PBD.

<table>
<thead>
<tr>
<th></th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>9</th>
<th>10</th>
<th>11</th>
</tr>
</thead>
<tbody>
<tr>
<td>(b_1)</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>(b_2)</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>(b_3)</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>(b_4)</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>(b_5)</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>(b_6)</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>(b_7)</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>(b_8)</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>(b_9)</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>(b_{10})</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>(b_{11})</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>(b_{12})</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>(b_{13})</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>(b_{14})</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>(b_{15})</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>(b_{16})</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
</tbody>
</table>

set of blocks.

We begin by removing variety \(n\) from all blocks in \(B\) and adding a single block \(\{n\}\). More precisely, let 
\[B' = \{b \setminus \{n\} | b \in B\} \cup \{\{n\}\}.\] Then 
\[B'\] contains \(n\frac{(n-1)}{6} - r\) (unchanged) blocks of weight 3, 
\[r = n - \frac{1}{2}\] blocks of weight 2, and a single block of 
weight 1.

Let \([S']\) be the incidence matrix of \((N, B')\). We will show that 
\[d([S']) \geq 3.\]

Our key observation comes from the fact that every pair of varieties in \(N\) 
appears in exactly one block in \(B\). Thus in \(B\), 
every pair of blocks differ by at least 4 elements. In terms of \([S]\), 
every pair of rows differs in at least four coordinates, meaning 
that the minimum distance of \([S]\) is at least 4. Because any two rows in \([S']\) of weight 3 correspond directly to 
blocks of \(S\), these two rows must have distance at least 4.

Let \(v_i\) and \(v_j\) be two rows of \([S']\) such that 
\[|v_i| = 3\] and \[|v_j| = 2.\] Based on a similar observation, vectors \(v_i\) and \(v_j\) share 
at most one 1 in the same coordinate. Thus, 
\[d(v_i, v_j) \geq 3.\]

Next consider two rows \(v_i\) and \(v_j\) of \([S']\) with 
\[|v_i| = 2\] and \[|v_j| = 2.\] By construction, the corresponding blocks 
\(b_i\) and \(b_j\) of \(S\) both originally contained the variety \(n\). Therefore 
\(b_i\) and \(b_j\) do not share any variety other than \(n\). Thus, \(v_i\) and \(v_j\) 
share 1’s in no coordinates. It follows that 
\[d(v_i, v_j) = 4.\] In particular, the rows of weight 2 in \([S']\) contain a single 1 in 
every column other than the column corresponding to variety \(n\), 
which has only 0’s.

Finally, because variety \(n\) has been removed from all blocks in \(B'\), 
the distance between any row \(v_i\) and the row \(v_n\) 
containing only variety \(n\) must be at least 3.

Thus, \(d([S']) \geq 3.\) In addition, \([S']\) contains one row 
with a weight of 1, \(n-1\) rows with weight 2, 
and the remaining \(n\frac{(n-1)}{6} - n - 1\) rows have weight 3, 
for a total of \(n\frac{(n-1)}{6} + 1\) rows. Thus \([S']\) forms a valid 
guessing game matrix for 
\[
\left(\frac{n(n-1)}{6} + 1\right)\text{ guessing game.}
\]

The construction above explains how to construct a matrix with exactly 
\[m = n\frac{(n-1)}{6} + 1\] rows. To create a valid matrix for 
a game with \(m < n\frac{(n-1)}{6} + 1\), we simply remove as many rows of \([S']\) as necessary, 
beginning with rows of weight 3. This 
cannot affect the minimum distance of the matrix.
Next, we consider a case for which a Steiner Triple System does not exist.

**Proposition 33** (See Lindner⁶). For \( n \equiv 5 \pmod{6} \), there exists an \((n, \{3, 5\}, 1)\)-PBD \( D \) such that \( D \) has exactly one block of size 5 while the rest of the blocks have size 3.

**Lemma 34.** Let \( n \equiv 5 \pmod{6} \) and \( 2 \leq m \leq \frac{n^2-n-2}{6} \). Suppose \( c = (1,1,\ldots,1) \). Then there exists an \( m \times n \) \((1,2,3)\) matrix \([G]\) which is a valid guessing game matrix for an \((m, c)\) guessing game \( G \) with exactly \( n \) questions.

**Proof.** Let \( n \equiv 5 \pmod{6} \). By Proposition 33, there exists a \((n, \{3, 5\}, 1)\)-PBD \( D \) with varieties \( N = \{1,2,\ldots,n\} \) and block set \( B \) that contains exactly one block of size 5 while the rest of the blocks have size 3. By counting pairs, we can determine that \( D \) contains \( \binom{\frac{n^2-n-2}{6}}{2} \) blocks of weight 3. Without loss of generality, we assume that \( \{n, b, c, d, e\} \) is the block of weight 5 in \( D \).

We now create a new block set by “breaking apart” the block of size 5. This will be similar to the proof of Lemma 32, but with the added complication of the block of size 5.

Let \( P = \{b \in B|n \in b \text{ and } |b| = 3\} \) be the set of triples in \( D \) containing the variety \( n \), and notice that \( |P| = \frac{n-5}{2} \). Define \( P' = \{b \setminus \{n\} | b \in P\} \). Let \( T = \{b \in B|n \notin b \text{ and } |b| = 3\} \) be the set of triples in \( D \) not containing the variety \( n \). We leave \( T \) unchanged.

We construct a new set of blocks \( B' = P' \cup T \cup \{(b, c), (d, e), \{n\}\} \). Note that \( \{(b, c), (d, e), \{n\}\} \) is essentially a “factoring” of the block of size 5. Then the set \( B' \) contains \( \binom{\frac{n^2-n-2}{6}}{2} + 3 = \frac{n^2-n-2}{6} + 3 \) blocks: \( \binom{\frac{n^2-n-2}{6}}{2} = \frac{n^2-n-2}{6} \) of weight 3 from \( T \), \( \frac{n-5}{2} \) of weight 2 from \( P' \), two new blocks also of weight 2, and the new block of weight 1. We note that \( B' \) contains \( \frac{n-5}{2} + 2 = \frac{n-1}{2} \) blocks of size 2, the maximum possible number of such blocks.

The resulting structure \( D' = (N, B') \) that has been constructed from \( D \) will have an incidence matrix \([D']\) in the form shown in Figure 14.

```
\begin{center}
\begin{tabular}{cccccc}
0 &  &  &  &  \\
1 & 1 & 0 & 0 & \cdots & 0 & 0 \\
0 & 0 & 1 & 1 & \cdots & 0 & 0 \\
 &  &  &  &  &  &  \\
0 & 0 & 0 & 0 & \cdots & 1 & 1 \\
0 & 0 & 0 & 0 & \cdots & 0 & 1 \\
\end{tabular}
\end{center}
```

**Figure 14.** The incidence matrix \([D']\).

Similarly to the proof of Lemma 32, by using the fact that every pair in \( N \) appears in exactly one block of \( D \), it follows that \( d([D']) \geq 3 \). Since we have shown that \([D']\) contains one row with a weight of \( 1, \frac{n-1}{2} \) rows with weight 2 which are mutually disjoint, while the remaining rows have weight 3 and \( d([D']) \geq 3 \), it follows that \([D']\) is a valid \((1,2,3)\) guessing game matrix for a \((\frac{n^2-n-2}{6}, c, 1)\) guessing game with \( n \) questions.

The construction above explains how to construct a matrix with exactly \( m = \frac{n^2-n-2}{6} \) rows. To create a valid matrix for a game with \( m < \frac{n^2-n-2}{6} \), we remove as many rows as necessary, beginning with rows of weight 3. This cannot affect the minimum distance of the matrix.

\( \square \)

Next, we will show that \((1,2,3)\) matrices (when they exist) give the lowest possible cost.
Theorem 35. Let \( c = (1, 1, \ldots, 1) \) and let \( n \geq 3 \). If a \( m \times n \) \((1, 2, 3)\) matrix \([M]\) exists, then \([M]\) attains the minimum total cost for an \((m, c, 1)\) guessing game with exactly \( n \) questions. This minimum cost is exactly \( 3(m - \lfloor \frac{n-1}{2} \rfloor - 1) + 2 \lfloor \frac{n-1}{2} \rfloor + 1 = 3m - \lfloor \frac{n-1}{2} \rfloor - 2 \).

Proof. Let \( m \) and \( n \) be parameters for an \((m, c, 1)\) guessing game with exactly \( n \) questions, and assume an \( m \times n \) \((1, 2, 3)\) matrix \([G]\) exists. Because the cost function is fixed at \( c = (1, 1, \ldots, 1) \), the total cost of a \((m, c, 1)\) guessing game with game matrix \([G]\) is exactly the number of 1’s in \([G]\). Thus our goal is to determine the minimum number of 1’s among all \( m \times n \) binary matrices \([G]\) which have a minimum distance of at least 3.

Let \([G]\) be a \( m \times n \) guessing game matrix with \( d([G]) \geq 3 \) that attains the minimum possible number of 1’s.

We first claim that \([G]\) does not contain a zero vector. If \([G]\) did contain a zero vector, then to ensure \( d([G]) \geq 3 \), all other vectors must have weight at least 3. Note that \([M]\) must contain at least one row of weight 2, and so this guarantees that \([G]\) has at least as many 1’s as \([M]\). Thus we may assume that \([G]\) does not contain a zero vector.

Next, we claim that \([G]\) must contain a single vector of weight 1. Certainly \([G]\) cannot have two or more vectors of weight 1 while maintaining \( d([G]) \geq 3 \), and so it can contain either 1 or 0 such vectors. Suppose it has none. Then even if \([G]\) has the maximum possible number of rows of weight 2, \([G]\) must have at least one more vector of weight 3 or larger compared to \([M]\). Thus \([G]\) would not have lower weight than \([M]\).

Thus we may assume that \([G]\) has no zero vector and has exactly one vector of weight 1. Without loss of generality, assume that this single 1 appears in column \( n \). To ensure \( d([G]) \geq 3 \), any rows of weight 2 may not have a 1 in column \( n \). Thus at most \( \lfloor \frac{n-1}{2} \rfloor \) of the remaining rows may have weight 2, and the rest must have weight at least 3. A \((1, 2, 3)\) matrix meets these bounds exactly, and so it must represent the minimum possible cost.

Note that the cost of an \( m \times n \) \((1, 2, 3)\) matrix is \( 3(m - \lfloor \frac{n-1}{2} \rfloor - 1) + 2 \lfloor \frac{n-1}{2} \rfloor + 1 \), because \( \lfloor \frac{n-1}{2} \rfloor \) of the \( m \) rows have weight 2, one has weight 1, and the rest of the \( m \) rows have weight 3.

\[ \square \]

Theorem 36. Let \( c = (1, 1, \ldots, 1) \) and let \( n \geq 3 \). Suppose that one of the following is true:

- \( n \equiv 1 \text{ or } 3 \pmod{6} \) and \( 2 \leq m \leq \frac{n(n-1)}{6} + 1 \)
- \( n \equiv 5 \pmod{6} \) and \( 2 \leq m \leq \frac{n^2-n}{6} - 2 \)

Then a \((1, 2, 3)\) matrix exists and provides the cheapest possible cost for an \((m, c, 1)\) guessing game with exactly \( n \) questions.

Proof. Under the assumptions on \( n \), an \( m \times n \) \((1, 2, 3)\) matrix \([M]\) exists by Lemmas 32 and 34. Note that if \( m \) is not as large as possible, then some rows of weight 3 must be removed from the corresponding \((1, 2, 3)\) matrix, as described in the proof of the respective Lemma. By Theorem 35, \([M]\) gives the cheapest possible cost for an \((m, c, 1)\) guessing game with exactly \( n \) questions.

\[ \square \]

We note that Theorem 36 covers the interesting case where \( m \leq \frac{n-1}{2} + 1 \), in which case a \((1, 2, 3)\) matrix \([M]\) contains only rows of weight 2 and 1 (because, per the definition, all rows of weight 3 must be removed before removing any rows of lower weight). In this case, \([M]\) is the same as a \((2, 1)\) matrix, although possibly with several extra columns containing all 0’s (effectively, questions that are unused). Thus Theorem 36 is consistent with the conclusion of Theorem 22 applied with \( c = (1, 1, \ldots, 1) \). In this case, the restriction on the number of questions has no effect.

We finish this section with an attempt to tie up the cases not considered by Theorem 36.

Definition 37. A punctured \((1, 2, 3)\) matrix has the same form as a \((1, 2, 3)\) matrix without the single row and column of weight one.

Theorem 38. Let \( c = (1, 1, \ldots, 1) \), \( n \) even with \( n \geq 2 \), and \( m \geq 2 \).

- If an \( m \times n \) \((1, 2, 3)\) matrix exists, then it attains the minimum possible cost among all \((m, c, 1)\) guessing games with exactly \( n \) questions.
• If an \( m \times n \) \((1, 2, 3)\) matrix does not exist, but an \( m \times n \) punctured \((1, 2, 3)\) matrix does exist, then the punctured \((1, 2, 3)\) matrix attains the minimum possible cost among all \((m, c, 1)\) guessing games with exactly \(n\) questions.

Proof. Recall that, because \(c = (1, 1, \ldots, 1)\), the cost of a game is exactly equal to the number of 1’s in that game’s matrix. If an \( m \times n \) \((1, 2, 3)\) matrix \([M]\) exists, then by Theorem 35 \([M]\) attains the minimum possible cost. Note that it is possible for an \( m \times n \) \((1, 2, 3)\) matrix to exist outside the conditions given in Lemmas 32 and 34.

If an \( m \times n \) \((1, 2, 3)\) matrix does not exist, then the minimum attainable cost is at least one greater than the cost that an \( m \times n \) \((1, 2, 3)\) matrix would give.

Suppose an \( m \times n \) \((1, 2, 3)\) matrix does not exist, but an \( m \times n \) punctured \((1, 2, 3)\) matrix \([M']\) does exist. The only difference in row weights between \([M']\) and a possible \(m \times n\) \((1, 2, 3)\) matrix is that \([M']\) will have a row of weight 2 where the \((1, 2, 3)\) matrix would have a row of weight 1, and thus \([M']\) would have a total cost of one more than the \(m \times n\) \((1, 2, 3)\) matrix. This is necessarily the minimum possible cost. \(\square\)

In particular, note Theorem 36 does not cover the cases where \(n \equiv 0, 2,\) or 4 \((\text{mod} \ 6)\). In these cases, \(n + 1 \equiv 1, 3\) or 5 \((\text{mod} \ 6)\). Thus if \(m + 1\) satisfies the hypotheses of Theorem 36, then a \((m + 1) \times (n + 1)\) \((1, 2, 3)\) matrix exists by Lemma 32 or 34. By removing the row and column of weight 1, we obtain a punctured \((1, 2, 3)\) matrix \([M']\) which is within 1 of the minimum possible cost by Theorem 38. However, we do not know whether or not a true \((1, 2, 3)\) matrix exists in these cases (and thus attains the minimum cost).

Example 39. The following is the construction of a \((1, 2, 3)\) matrix with \(m = 8\) as described in this section. We begin with the incidence matrix of an STS(7) from Figure 12.

By following the construction of a \((1, 2, 3)\) matrix as described in Lemma 32, we let the variety 1 be the variety that we remove from the blocks and then replace in a block of weight 1. From this, we yield the following matrix.

<table>
<thead>
<tr>
<th></th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
</tr>
</thead>
<tbody>
<tr>
<td>b1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>b2</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>b3</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>b4</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>b5</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>b6</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>b7</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
</tr>
<tr>
<td>b8</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
</tbody>
</table>

Figure 15. The incidence matrix with variety 1 removed, and a new row added.

By rearranging the columns and rows of the matrix in Figure 15, we obtain the \((1, 2, 3)\) matrix in Figure 16.

If \(c = (1, 1, 1, 1, 1, 1)\), this gives a \((8, c, 1)\) guessing game with exactly 7 questions and total cost \(3 \cdot 8 - \lfloor \frac{8 - 1}{2} \rfloor - 2 = 19\).

Example 40. The following is the construction of a \((1, 2, 3)\) matrix with \(m = 11\) as described in this section. We begin with the incidence matrix of an \((11,\ (3, 5, 1)\)-PBD from Figure 13.

By following the construction of a \((1, 2, 3)\) matrix as described in Lemma 34, we let the variety 1 be the variety that we remove from the blocks and then replace in a block of weight 1. From this, we yield the matrix seen in Figure 17. The blocks created from original block of weight 5 are highlighted.

By rearranging the columns and rows of the matrix in Figure 17, we obtain the \((1, 2, 3)\) matrix in Figure 18.
\begin{figure}
\begin{center}
\begin{tabular}{cccccccc}
& 2 & 4 & 3 & 7 & 5 & 6 & 1 \\
\hline
\(b_2\) & 1 & 0 & 1 & 0 & 1 & 0 & 0 \\
\(b_3\) & 0 & 1 & 1 & 0 & 0 & 1 & 0 \\
\(b_4\) & 0 & 1 & 0 & 1 & 1 & 0 & 0 \\
\(b_6\) & 1 & 0 & 0 & 1 & 0 & 1 & 0 \\
\(b_1\) & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\
\(b_7\) & 0 & 0 & 1 & 1 & 0 & 0 & 0 \\
\(b_5\) & 0 & 0 & 0 & 0 & 1 & 1 & 0 \\
\(b_8\) & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\
\end{tabular}
\caption{The final \((1, 2, 3)\) matrix.}
\end{center}
\end{figure}

\begin{figure}
\begin{center}
\begin{tabular}{cccccccccc}
& 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 \\
\hline
\(b_1\) & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
\(b_2\) & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\
\(b_3\) & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\
\(b_4\) & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 0 \\
\(b_5\) & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\
\(b_6\) & 0 & 1 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 \\
\(b_7\) & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 \\
\(b_8\) & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 \\
\(b_9\) & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\
\(b_{10}\) & 0 & 0 & 1 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 \\
\(b_{11}\) & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 \\
\(b_{12}\) & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 1 & 0 \\
\(b_{13}\) & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 1 & 0 & 0 & 0 \\
\(b_{14}\) & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 1 & 0 & 1 \\
\(b_{15}\) & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 0 & 1 \\
\(b_{16}\) & 0 & 0 & 0 & 0 & 1 & 0 & 1 & 0 & 0 & 1 & 0 \\
\(b_{17}\) & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 1 & 1 & 0 & 0 \\
\(b_{18}\) & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
\end{tabular}
\caption{}
\end{center}
\end{figure}

\section*{ACKNOWLEDGEMENTS}
Thank you to the Student Summer Scholars (S3) program and at Grand Valley State University and all of the program’s wonderful donors for funding this research. Another thank you goes out to my research advisor, Dr. David Clark, for his guidance, advice, and positivity throughout this project.

\section*{REFERENCES}
<table>
<thead>
<tr>
<th></th>
<th>2</th>
<th>3</th>
<th>4</th>
<th>5</th>
<th>6</th>
<th>7</th>
<th>8</th>
<th>11</th>
<th>9</th>
<th>10</th>
<th>1</th>
</tr>
</thead>
<tbody>
<tr>
<td>$b_6$</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_7$</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_8$</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>$b_9$</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_{10}$</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_{11}$</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>$b_{12}$</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>$b_{13}$</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_{14}$</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_{15}$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_{16}$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
</tr>
<tr>
<td>$b_{17}$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_1$</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_2$</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_3$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_4$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_5$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
<td>1</td>
<td>0</td>
<td>0</td>
</tr>
<tr>
<td>$b_{18}$</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>1</td>
</tr>
</tbody>
</table>

Figure 18. The final $(1, 2, 3)$ matrix.


ABOUT THE STUDENT AUTHOR
Lindsay Czap is a senior mathematics student at Grand Valley State University. At Grand Valley, she participated in the Student Summer Scholars program which funded this project. She will be continuing her mathematics education as she pursues a Master’s degree at Bowling Green State University.

PRESS SUMMARY
The game “20 questions” has always been a staple for children trapped in long car rides. We explore guessing games like “20 questions” in which one player chooses a number, and the other player asks yes-no questions to determine the number. In our games, the responding player may lie once. By adding a cost to each question in these games, we are able to quantify their efficiency. We determine the most efficient games in two cases: When the number of questions is unlimited, and when the number of questions is limited and the costs are especially simple. In each case, we explain exactly how to play these games in the most efficient way.