2-person 0-sum (or constant sum) game theory
Quotes of the Day

My work is a game, a very serious game.
   -- M. C. Escher (1898 - 1972)

Conceal a flaw, and the world will imagine the worst.
   -- Marcus Valerius Martialis (40 AD - 103 AD)

Reveal your strategy in a game, and your outcome will be the worst.
   -- Professor Orlin (for the lecture on game theory)
We will cover 2-person 0-sum game theory for several reasons:

1. It is closely connected to linear programming and duality
2. It introduces some important concepts in game theory
3. I think it is fun.
I read Marilyn Vos Savant’s column on a regular basis. Some of my favorite columns of hers are ones that deal with elementary probability or mathematics.
2-person 0-sum Game Theory

- Two people make decisions at the same time.
- The payoff depends on the joint decisions.
- Whatever one person wins the other person loses.
  - Marilyn vos Savant answered the question incorrectly. [http://www.siam.org/siamnews/06-03/gametheory.pdf](http://www.siam.org/siamnews/06-03/gametheory.pdf)
Payoff Matrix for Vos Savant’s Game

You choose heads or tails
The beautiful stranger chooses heads or tails

<table>
<thead>
<tr>
<th></th>
<th>Heads</th>
<th>Tails</th>
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</thead>
<tbody>
<tr>
<td><strong>Heads</strong></td>
<td>3</td>
<td>-2</td>
</tr>
<tr>
<td><strong>Tails</strong></td>
<td>-2</td>
<td>1</td>
</tr>
</tbody>
</table>

This matrix is the payoff matrix for you, and the beautiful stranger gets the negative.
Payoff Matrix for Rock-Paper-Scissors

Person R chooses a row: either 1, 2, or 3

Person C chooses a column: either 1, 2, or 3

<table>
<thead>
<tr>
<th></th>
<th>Rock</th>
<th>Paper</th>
<th>Scissors</th>
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</thead>
<tbody>
<tr>
<td>Rock</td>
<td>0</td>
<td>-1</td>
<td>1</td>
</tr>
<tr>
<td>Paper</td>
<td>1</td>
<td>0</td>
<td>-1</td>
</tr>
<tr>
<td>Scissors</td>
<td>-1</td>
<td>1</td>
<td>0</td>
</tr>
</tbody>
</table>

This matrix is the payoff matrix for player R. (And player C gets the negative.)
Another Payoff Matrix

Person R chooses a row: either 1, 2, or 3

Person C chooses a column: either 1, 2, or 3

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<thead>
<tr>
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<th>1</th>
<th>2</th>
<th>3</th>
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<tbody>
<tr>
<td>1</td>
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<td>2</td>
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<tr>
<td>3</td>
<td>1</td>
<td>0</td>
<td>-2</td>
</tr>
</tbody>
</table>

This matrix is the payoff matrix for player R. (And player C gets the negative.)

e.g., R chooses row 3; C chooses column 1

R gets 1; C gets −1 (zero sum)
Some more examples of payoffs

R chooses 2, C chooses 3
R gets 0; C gets 0 (zero sum)

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<thead>
<tr>
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<tbody>
<tr>
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<tr>
<td>1</td>
<td></td>
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</tbody>
</table>

R chooses row 3; C chooses column 3
R gets -2; C gets +2 (zero sum)
Next: 2 volunteers

Player R puts out 1, 2 or 3 fingers

Player C simultaneously puts out 1, 2, or 3 fingers

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<thead>
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<td>2</td>
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</table>

R tries to maximize his or her total
C tries to minimize R’s total.

We will run the game for 5 trials.
Next: Play the game with your partner
(If you don’t have one, then watch)

Player R puts out 1, 2 or 3 fingers

Player C simultaneously puts out 1, 2, or 3 fingers

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<thead>
<tr>
<th></th>
<th>R</th>
<th>C</th>
<th>Tie</th>
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<tbody>
<tr>
<td>-2</td>
<td>1</td>
<td>2</td>
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<tr>
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<tr>
<td>1</td>
<td>0</td>
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<td></td>
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</tbody>
</table>

R tries to maximize his or her total
C tries to minimize R’s total.

Run the game for 5 trials.

Total
Who has the advantage: R or C?

Here we will take a very conservative viewpoint?
What can player R guarantee in return, regardless of what C chooses?

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<thead>
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<th>1</th>
<th>2</th>
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<tbody>
<tr>
<td>2</td>
<td>-1</td>
<td>0</td>
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<tr>
<td>1</td>
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Most students guess correctly that R has an advantage. But to determine who has an advantage in general is a much more difficult problem than it seems, and relies on linear programming.
Can R guarantee an expected return independent of what C does? (a lower bound on the opt for R).

Suppose R chooses row j. What can R guarantee?

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<thead>
<tr>
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</thead>
<tbody>
<tr>
<td>2</td>
<td>-1</td>
<td>0</td>
</tr>
<tr>
<td>1</td>
<td>0</td>
<td>-2</td>
</tr>
</tbody>
</table>

What row offers R the best guaranteed “return”?

A strategy that consists of selecting the same row over and over again is a “pure strategy.” R can guarantee a payoff of at least –1.

What column offers C the best guaranteed “return”? 
Random (mixed) strategies

Suppose we permit R to choose a random strategy. What can R guarantee.

Suppose R will flip a coin, and chooses: Row 1 if Heads, and Row 3 if tails. That is, .5 for Row 1 .5 for Row 3

The idea of a randomized strategy is one of the most important concepts of game theory. A good example occurs in sporting events, such as baseball. If a batter knows what a pitcher is going to throw, then the batter has an advantage. So the pitcher will try to “mix it up” so that the batter cannot know for sure what kind of pitch he will receive. In fact, it is optimal for the pitcher to randomize.

Similarly, the batter often has to decide in advance what kind of pitch to swing at. If he were to let the pitcher know what he is expecting, then the batter is at a disadvantage. So, the batter is better off randomizing on what he will be looking for.
So, with a random strategy R guarantees at least -.5 regardless of what column C chooses.

That is, the minimum expected payoff over all three columns is -.5.

In two person game theory, we will often talk in terms of guarantees. Of course, the guarantee is not a real guarantee in the usual sense. It is a guarantee in terms of expected value. If the row player chose this strategy a large number of times, then he or she would receive at least -.5 on average.

This is not a particularly good guarantee, but it is better than the -1 from before. We shall see that the row player can do much better.
Another example: Player R chooses each row with probability of 1/3.

<table>
<thead>
<tr>
<th>Prob.</th>
<th>-2</th>
<th>1</th>
<th>2</th>
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<tbody>
<tr>
<td>1/3</td>
<td></td>
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</tr>
<tr>
<td>1/3</td>
<td>2</td>
<td>-1</td>
<td>0</td>
</tr>
<tr>
<td>1/3</td>
<td>1</td>
<td>0</td>
<td>-2</td>
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</tbody>
</table>

Expected Payoff

Thus R can guarantee a payoff of at least _____.

It is a lower bound on what R will get.

This is better. With this simple mixed strategy, the row player can guarantee at least 0 regardless of what the column player does. Moreover, if the column player were to occasionally choose column 1, then the row player would do even better.
Player C can also do this. Suppose that Player C chooses each column with probability of 1/3.

<table>
<thead>
<tr>
<th>Prob.</th>
<th>1/3</th>
<th>1/3</th>
<th>1/3</th>
<th>Expected Payoff</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Row 1</td>
</tr>
<tr>
<td>-2</td>
<td>1</td>
<td>2</td>
<td></td>
<td></td>
</tr>
<tr>
<td>2</td>
<td>-1</td>
<td>0</td>
<td></td>
<td></td>
</tr>
<tr>
<td>1</td>
<td>0</td>
<td>-2</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Row 2</td>
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<td></td>
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<tr>
<td></td>
<td></td>
<td></td>
<td></td>
<td>Row 3</td>
</tr>
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<td></td>
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</tr>
</tbody>
</table>

Thus C can limit R’s payoff to at most _____.

This is the maximum expected payoff over all three rows

This gives an upper bound on what R would get.

Previously we were focusing on guarantees that the row player can achieve. Of course, this is a two person game, and so we can carry out similar analysis for the column player. In this case, the column player can guarantee that the row player receives at most 1/3, which is what would happen if the row player selected row 1 or row 2. The payoff to the row player could be less than 1/3 in this case if the row player occasionally chose row 3.
Challenge to Class

\[
\begin{array}{ccc}
-2 & 1 & 2 \\
2 & -1 & 0 \\
1 & 0 & -2 \\
\end{array}
\]

Left half of class: row players. Choose probabilities for each row so that you are guaranteed more than 0, regardless of what the column player does.

Right half of class: column players. Choose probabilities for each column so that the row player can get no more than .3 regardless of what the row player does.

Even though this is a tiny problem, it is extremely hard to come up with the optimum probabilities for the row player or for the column player. It is a challenge enough to do better than 0 for the row player or for the column player to guarantee a payoff to R of less than .3.
Row Player’s Problem

Let z be a lower bound on what R will get.

Maximize $z = \min\{-2x_1 + 2x_2 + x_3, x_1 - x_2, 2x_1 - 2x_3\}$

s.t. $x_1 + x_2 + x_3 = 1$

$x_1, x_2, x_3 \geq 0$

This is the maximin payoff.

We finally see the connection between two person 0-sum games and optimization. The Row Player’s problem is to develop probabilities so that the guaranteed payoff is as large as possible. The column player can choose column 1 or column 2 or column 3, each leading to its own expected payoff. The guaranteed payoff is the least of these three expected payoffs. And we need $x_1, x_2,$ and $x_3$ to correspond to probabilities. So, they are nonnegative and they sum to 1.

We call this the maximin payoff because we are maximizing the minimum of three payoff values.
Row Player’s Problem

Maximize $z = \min\{ -2x_1 + 2x_2 + x_3, \]
\[ \quad x_1 - x_2, \]
\[ \quad 2x_1 - 2x_3 \}
\]
s.t. $x_1 + x_2 + x_3 = 1$
\[ x_1, x_2, x_3 \geq 0 \]

Recall:
$\min(1, 0, 2)$
is the same as
\[ \max z \]
s.t. $z \leq 1$
\[ z \leq 0 \]
\[ z \leq 2 \]

In Lecture 2 of 15.053, we learned how to convert a maximin problem into a linear program. We carry out the same transformation here. So, the optimum strategy for the row player (in the maximin sense) can be determined by solving a linear program.
This is what happens in 2-person 0-sum games in general. Instead of a 3 by 3 payoff matrix, there is an \( m \) by \( n \) payoff matrix. The row player will select probabilities for \( m \) rows. This vector \( x \) of probabilities leads to a payoff for each column. The row player wants to choose \( x \) so that the minimum payoff of a row is maximized.
A similar analysis can be carried out for the column player. In this case, the column players chooses probabilities associated with each column.

If the row player chooses row 1, then the associated payoff is $-2y_1 + y_2 + 2y_3$. There are also payoffs for the other two rows. The column player wants to minimize the guaranteed payoff to the row player which is the max of those three payoffs. For this reason, the column player’s problem is a minimax problem.
Similar to before, we translate the column player’s problem into a linear program.
The optimal strategy for R (in terms of maximin optimization) is to choose Row one 7/18 of the time, to choose row two 5/18 of the time, and to choose row three 1/3 of the time. Under these circumstances, the row player will receive an expected payoff of 1/9 regardless of what the column player does.

A natural question is: does the optimal strategy for the row player always lead to a payoff that is the same for all columns. The answer is no. For example, if you added a fourth column of (0, 0, .5), you would find that the expected payoff of that column is 1/6. And it would also be true that the optimum strategy for the row player would be the same as above, with an expected maximin payoff of 1/9. The difference in this case is that the row player could possibly get more than 1/9 if the column player occasionally selected the fourth column.
An *optimal* randomized strategy for C.

<table>
<thead>
<tr>
<th>Prob.</th>
<th>1/3</th>
<th>5/9</th>
<th>1/9</th>
<th>Exp. payoff</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>-2</td>
<td>1</td>
<td>2</td>
<td>1/9</td>
</tr>
<tr>
<td></td>
<td>2</td>
<td>-1</td>
<td>0</td>
<td>1/9</td>
</tr>
<tr>
<td></td>
<td>1</td>
<td>0</td>
<td>-2</td>
<td>1/9</td>
</tr>
</tbody>
</table>

So, if C plays the optimal minimax strategy, then R gets at most 1/9.

If we carry out the analysis for the column player, the optimal minimax strategy is for the column player to choose column one with probability 1/3, column two with probability 5/9, and column three with probability 1/9. In this case, the column player can guarantee a payoff of at most 1/9 to the row player regardless of what row the row player selects.

It turns out that the maximin payoff is the same as the minimax payoff, which is remarkable. After all, the maximin payoff is a very conservative strategy. Rather than trying to take advantage of knowledge of his or her opponent, the row player chooses probabilities that are absolutely guaranteed to return an expected payoff of 1/9 regardless of what the column player does. It seems plausible that if one takes into account lots of information about the column player, possibly the row player would do even better.

But no. The row player cannot do better if the column player is also playing conservatively. In this case, the column player can guarantee a payoff of at most 1/9 to the row player, regardless of what the row player does.

To see how extreme and unintuitive this result is, imagine that you are the row player and you have to announce in advance of playing what your mixed (randomized) strategy is. You may think that announcing your strategy prior to the column player selecting a column has to be a bad idea, and will cost you. But if you announce the optimum maximin strategy, you will still guarantee an expected payoff of 1/9. Announcing the strategy does not cost you int his case.
Fundamental Theorem of 2-person 0-sum Game Theory.

For any two person game, let RPay(x) be the guaranteed payoff to R if the row player chooses probability vector x. Let CPay(y) be the upper bound on what R can get if the column player chooses probability vector y. Then

\[ \text{RPay}(x) \leq \text{CPay}(y) \]  

“Weak Duality”

There exists x* and y* so that

\[ \text{RPay}(x^*) = \text{CPay}(y^*) \]  

“Strong Duality”

That is, maximin payoff = minimax payoff.

Proof: The column player’s LP is the dual of the row player’s LP.
The two linear programs given before do not seem to be duals of each other. But if we rewrite them, we can see that they are duals. Using the SOB technique, one can easily see that the second LP is the dual of the first.
Developers of Game Theory

John Von Neumann  Oskar Morgenstern
Class Break

Von Neumann

RPS
From Marilyn Vos Savant’s column.

Say you’re in a public library, and a beautiful stranger strikes up a conversation with you. She says: ‘Let’s show pennies to each other, either heads or tails. If we both show heads, I pay you $3. If we both show tails, I pay you $1. If they don’t match, you pay me $2.’

At this point, she is shushed. You think: ‘With both heads 1/4 of the time, I get $3. And with both tails 1/4 of the time, I get $1. So 1/2 of the time, I get $4. And with no matches 1/2 of the time, she gets $4. So it’s a fair game.’ As the game is quiet, you can play in the library.

But should you? Should she?

We now return to Marilyn vos Savant’s problem.
Determining the optimal strategy

Choose the value of $p$ that maximizes the minimum payoff.

<table>
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<tr>
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<th>B. S.</th>
<th>H</th>
<th>T</th>
<th>Prob</th>
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<tbody>
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<td>3</td>
<td>-2</td>
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<td>$p$</td>
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<tr>
<td>T</td>
<td>-2</td>
<td>1</td>
<td></td>
<td>$1-p$</td>
</tr>
</tbody>
</table>

Col 1  Col 2

$maximize \quad z = \min \{3p - 2(1-p), -2p + 1(1-p) \}

subject to \quad 0 \leq p \leq 1$

Col 1  Col 2

$maximize \quad z = \min \{5p - 2, -3p + 1 \}

subject to \quad 0 \leq p \leq 1$

Note that the row player only has to choose one variable, that is, $p$. If we write the maximin problem, it is a problem involving two variables: $p$ and $z$. We can express this as a linear program, but we will leave it in the maximin form instead.

The second optimization problem is a simplification of the first one.
Determining the optimal strategy

\[
\begin{align*}
\text{Col 1} & \quad \text{Col 2} \\
\text{maximize} & \quad z = \min \{5p - 2, -3p + 1\} \\
\text{subject to} & \quad 0 \leq p \leq 1
\end{align*}
\]

If we write the min of two linear functions (as \( p \) varies), we will see that it is the blue line given in the graph. If we then find the maximum of the blue line, it is the black dot.
The beautiful stranger’s viewpoint

Choose the value of $y$ that minimizes that maximum payoff.

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<thead>
<tr>
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<th>B. S.</th>
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<tbody>
<tr>
<td>H</td>
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<table>
<thead>
<tr>
<th></th>
<th>1-y</th>
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<tbody>
<tr>
<td>Prob</td>
<td>y</td>
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</tbody>
</table>

Row 1 Row 2

minimize $z = \max \{3y + -2(1-y), -2y + 1(1-y) \}$
subject to $0 \leq y \leq 1$

Row 1 Row 2

minimize $z = \max \{5y – 2, -3y + 1 \}$
subject to $0 \leq y \leq 1$

We can carry out the same analysis from the beautiful stranger’s viewpoint. In this case, we let $y$ be the probability of the column player choosing heads.
The payoff as a function of $y$ is the max of two linear functions, which is represented as the blue line.

The beautiful stranger chooses the minimum of the blue line, which is at the black dot. It turns out that this is when $y = 3/8$. The payoff to the row player (that is, you) is $-1/8$, and thus the beautiful stranger wins.
The payoffs are the same when $y = 3/8$

Optimal payoff to row player = $-1/8$

Marilyn vos Savant chose $y = 1/3$, which would give the B.S. a payoff of 0.

But inexplicably, Marilyn vos Savant chose $y = 1/3$, which is not optimal.
Summary

- 2-person 0 sum game theory
- Can be solved using LP
- 2-dimensional version can be solved graphically
- Fundamental theorem
- Marilyn vos Savant may be the person with the highest measured IQ, but she does not seem to know her game theory.