

# Notebook

August 10, 2017

```
In [1]: import numpy as np
import nash
```

## 1 Player strategies

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### 1.1 Definition of mixed strategies

A mixed strategy for a player with strategy set  $S$  is denoted by  $\sigma \in [0, 1]_{\mathbb{R}}^{|S|}$  and corresponds to a probability distribution over the pure strategies of player  $i$ . So:

$$\sum_{i=1}^{|S|} \sigma_i = 1$$

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The expected score of a player can then be calculated as a measure over the probability distributions.

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### 1.2 Calculating utilities

Considering a game  $(A, B) \in \mathbb{R}^{\times}$ , if  $\sigma_r$  and  $\sigma_c$  are the mixed strategies for the row / column player (respectively). The utility to the row player is:

$$u_r(\sigma_r, \sigma_c) = \sum_{i=1}^m \sum_{j=1}^n A_{ij} \sigma_{ri} \sigma_{cj}$$

and the utility to the column player is:

$$u_c(\sigma_r, \sigma_c) = \sum_{i=1}^m \sum_{j=1}^n B_{ij} \sigma_{ri} \sigma_{cj}$$

This comes from:

- The probability of being in a given cell of  $A$  or  $B$ :  $\sigma_{ri} \sigma_{cj}$
- The value of the particular cell:  $A_{ij}$  or  $B_{ij}$

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As an example consider the matching pennies game:

$$A = \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \quad B = \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix}$$

with the following mixed strategies:

$$\sigma_r = (.2, .8) \quad \sigma_c = (.6, .4)$$

We have:

$$u_r(\sigma_r, \sigma_c) = 0.2 \times 0.6 \times 1 + 0.2 \times 0.4 \times (-1) + 0.8 \times 0.6 \times (-1) + 0.8 \times 0.4 \times 1 = -0.12$$

$$u_c(\sigma_r, \sigma_c) = 0.2 \times 0.6 \times (-1) + 0.2 \times 0.4 \times 1 + 0.8 \times 0.6 \times 1 + 0.8 \times 0.4 \times (-1) = 0.12$$

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### 1.3 Linear algebraic calculation

Note that we can rearrange the expressions for the utilities:

$$u_r(\sigma_r, \sigma_c) = \sum_{i=1}^m \sigma_{ri} \sum_{j=1}^n A_{ij} \sigma_{cj}$$
$$u_c(\sigma_r, \sigma_c) = \sum_{i=1}^m \sigma_{ri} \sum_{j=1}^n B_{ij} \sigma_{cj}$$

in turn this corresponds to the matrix vector product:

$$u_r(\sigma_r, \sigma_c) = \sigma_r A \sigma_c^T$$

$$u_c(\sigma_r, \sigma_c) = \sigma_r B \sigma_c^T$$

We can use numpy to verify this calculation:

```
In [2]: A = np.array([[1, -1], [-1, 1]])
        B = np.array([[-1, 1], [1, -1]])
        sigma_r = np.array([.2, .8])
        sigma_c = np.array([.6, .4])
        np.dot(sigma_r, np.dot(A, sigma_c)), np.dot(sigma_r, np.dot(B, sigma_c))
```

```
Out[2]: (-0.11999999999999998, 0.11999999999999998)
```

Finally we can also directly calculate this using a nashpy game:

```
In [4]: matching_pennies = nash.Game(A, B)
        matching_pennies[sigma_r, sigma_c]
```

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Out[4]: array([-0.12,  0.12])
```