## Predicting a tennis match in progress for sports multimedia

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#### **Abstract**

This paper demonstrates how spreadsheets can generate the probability of winning a tennis match conditional on the state of the match. Previous models treat games, sets and matches independently. We show how a series of interconnected sheets can be used to repeat these results. The sheets are used in multimedia to predict outcomes for a match in progress, where it is shown how these predictions could benefit the spectator, punter, player and commentator. The development of the predictions could also form an interesting and useful teaching example, and allow students to investigate the properties of tennis scoring systems.

Keywords: Markov processes, Sports, Probability, Forecasting

#### 1. Introduction

Profiting from sports betting is an obvious application of predicting outcomes in sport. Whilst sports betting was originally restricted to betting prior to the start of the match, it is now possible and common to be betting throughout a match in progress. However, the appeal of predictions throughout a match may not just be restricted to punters. In-play sports predictions could be used in sports multimedia, and hence could be appealing to the spectator, coach or technology buff; without involving actual betting.

According to Wikipedia: "Multimedia is media and content that includes a combination of text, audio, still images, animation, video, and interactivity content forms. Multimedia is usually recorded and played, displayed or accessed by information content processing devices, such as computerized and electronic devices". Cadability Pty. Ltd. is a sports multimedia organization that specializes in delivering online information during live matches. As well as displaying the scoreboard, live predictions are in operation for tennis, cricket, Australian Rules football and soccer. This information is available on a standard PC and iPhone (as an iPhone app), with current developments for delivering the information on an iPad and TV (in the form of a widget). There are many ways the predictions through multimedia could be used. Spectators could engage with live predictions through multimedia for entertainment when watching a live match. If a spectator was to place a bet, then live predictions through multimedia could be used as a decision

support tool as to when and how much to bet on a particular event, and hence the combination of betting and multimedia becomes a powerful form of entertainment.

Another interesting application of sports predictions in multimedia is in teaching mathematical concepts. Students often relate to sporting events and hence may be stimulated in learning mathematics through an activity of personal interest. In tennis for example, the chance of winning from deuce is calculated by the sum of an infinite series. Another application is in using the predictions as a coaching tool. For example, it is common for players and coaches to watch a replayed match to discuss strategies for upcoming matches. The graphical and visual aspects of the predictions could enhance the TV replay. A further application is in using the predictions for TV commentary. Klaassen and Magnus (2003) demonstrate how plotting points on a graph on the probability of winning a tennis match in progress can be useful for TV commentary by supporting his/her discussion on the likely winner of the match. Commentators could also use the graph to evaluate the match after completion to identify turning points and key shifts in momentum.

In this paper, a tennis prediction model during a match in progress is constructed. Data analysis is carried out in section 2 to determine the parameters for the model. A Markov chain model is then developed in section 3 and an updating rule is given in section 4 to update prior estimates with what has occurred during the match. The men's 2010 US Open tennis final is given in section 5 as an illustration of live sports predictions through multimedia.

## 2. Data Analysis

By assigning two parameters, the constant probabilities of player A and player B winning a point on serve; modelling the probability of winning the match can be determined using a Markov Chain model as represented in section 3. This section will therefore derive the probabilities of winning on serve when two players meet on a particular surface. This is achieved by collecting, combining and updating player serving and receiving statistics.

## 2.1 Collecting player statistics

OnCourt (www.oncourt.info) is a software package for all tennis fans, containing match results for men's and woman's tennis, along with statistical information about players, tournaments or histories of the head-to-head matches between two players. Match statistics can be obtained for the majority of ATP and WTA matches going back to 2003. The Association of Tennis Professionals (ATP) is the governing body of men's professional tennis for the allocation of player rating points in matches to determine overall rankings and seedings for tournaments. The Women's Tennis Association (WTA) is used similarly in women's professional tennis. Table 1 gives the match statistics broadcast from the US Open 2010 men's final where Rafael Nadal

defeated Novak Djokovic in four sets. Notice that the Serving Points Won is not given directly in the table. This statistic can be derived from the Receiving Points Won such that Serving Points Won for Nadal and Djokovic are 1-36/112 = 67.9% and 1-60/143 = 58.0% respectively. Alternatively, the Serving Points Won can be obtained from a combination of the 1<sup>st</sup> Serve %, Winning % on 1<sup>st</sup> Serve and Winning % on 2<sup>nd</sup> Serve such that Serving Points Won for Nadal and Djokovic are 75/112\*55/75+(1-75/112)\*21/37= 67.9% and 95/143\*61/95+(1-95/143)\*22/48 = 58.0% respectively. Note that the Winning % on 1<sup>st</sup> Serve is conditional on the 1<sup>st</sup> Serve going in whereas the Winning % on the 2<sup>nd</sup> Serve is unconditional on the 2<sup>nd</sup> Serve going in. These calculations could be used as a teaching exercise in interpreting and analyzing data, and in conditional probabilities. Many more calculations can be obtained from broadcast match statistics as outlined in Bedford et al. (2010).

	Rafael Nadal	Novak Djokovic
1st Serve %	75 of 112 = 66%	95 of 143 = 66%
Aces	8	5
Double Faults	2	4
Unforced Errors	31	47
Winning % on 1 <sup>st</sup> Serve	55 of 75 = 73%	61 of 95 = 64%
Winning % on 2 <sup>nd</sup> Serve	21 of 37 = 56%	22 of 48 = 45%
Winners (Including Service)	49	45
Break Point Conversions	6 of 26 = 23%	3 of 4 = 75%
Receiving Points Won	60 of 143 = 41%	36 of 112 = 32%
Net Approaches	16 of 20 = 80%	28 of 45 = 62%
Total Points Won	136	119
Fastest Serve	212 KPH	201 KPH
Average 1 <sup>st</sup> Serve Speed	186 KPH	188 KPH
Average 2 <sup>nd</sup> Serve Speed	141 KPH	151 KPH

Table 1: Match statistics for the men's 2010 US Open final between Rafael Nadal and Novak Djokovic

### 2.2 Combining player statistics

Combining player statistics is a common challenge in sport. While we would expect a good server to win a higher proportion of serves than average, this proportion would be reduced somewhat if his opponent is a good receiver. Using the method developed by Barnett and Clarke (2005) we can calculate the percentage of points won on serve when player i meets player j on surface s ( $f_{ijs}$ ) as:

$$f_{ijs} = f_{is} - g_{js} + g_{avs} \tag{1}$$

where:

 $f_{is}$  = percentage of points won on serve for player i on surface s,

 $g_{is}$  = percentage of points won on return of serve for player i on surface s

 $g_{avs}$  represents the average (across all ATP/WTA players) percentage of points won on return of serve on surface s.

The surfaces are defined as: s=1 for grass, s=2 for carpet, s=3 for hard and s=4 for clay.

The average percentage of points won on serve across all players on each of six different surfaces (grass, hard, indoor hard, clay, carpet and acrylic) was calculated from OnCourt and represented in table 2. Note that the serving averages for carpet and indoor hard are approximately the same and are therefore combined as the one surface. Similarly hard and acrylic are combined as the one surface.

Example: Suppose player i with  $f_{i1}$ = 0.7 and  $g_{i1}$ = 0.4 meets player j with  $f_{j1}$ = 0.68 and  $g_{j1}$ = 0.35 on a grass court surface. Then the estimated percentage of points won on serve for player i and player j are given by  $f_{ij1}$  = 0.7-0.35+(1-0.653) = 69.7% and  $f_{ji1}$  = 0.68-0.4+(1-0.653)=62.7% respectively.

Surface	Men	Women
Grass	0.653	0.580
Carpet – I.hard	0.642	0.570
Hard – Acrylic	0.625	0.552
Clay	0.600	0.536

Table 2: The average probabilities of points won on serve for men's and women's tennis

### 2.3 Updating player statistics

The general form for updating the rating of a player as given by Clarke (1994) is New Rating = Old Rating +  $\alpha$  [actual margin – predicted margin]

for some  $\alpha$ .

Using serving and receiving player statistics as ratings we get

$$f_{is}^{\ n} = f_{is}^{\ o} + \alpha_s [f_{is}^{\ a} - f_{ijs}]$$
 (2)

$$g_{is}^{n} = g_{is}^{o} + \alpha_{s} [g_{is}^{a} - g_{ijs}]$$
 (3)

where.

 $f_{is}^{\ n}$ ,  $f_{is}^{\ o}$  and  $f_{is}^{\ a}$  represent the new, old and actual percentage of points won on serve for player i on surface s

 $g_{is}^{\ n}$ ,  $g_{is}^{\ o}$  and  $g_{is}^{\ a}$  represent the new, old and actual percentage of points won on return of serve for player i on surface s

 $\alpha_{s}\, is$  the weighting parameter for surface s

Experimental results reveal that  $\alpha_s = 0.049$  is a suitable weighting parameter for all surfaces. Further, every player is initialized with surface averages as given in table 2.

Equations (2) and (3) treat each surface independently. A more advanced approach is to update the serving and receiving statistics for each surface when playing on a particular surface. For example, if a match is played on grass, then how are the other surfaces of clay, carpet and hard court updated based on the player's performances on the grass?

This more complicated approach is given as:

$$f_{ist}^{\ n} = f_{is}^{\ o} + \alpha_{st} [f_{is}^{\ a} - f_{ijs}]$$
 (4)

$$g_{ist}^{n} = g_{is}^{o} + \alpha_{st} [g_{is}^{a} - g_{ijs}]$$
 (5)

where:

 $f_{ist}^{n}$  represents the new expected percentage of points won on serve for player i on surface t when the actual match is played on surface s

 $g_{ist}^{\ n}$  represents the new expected percentage of points won on return of serve for player i on surface t when the actual match is played on surface s

 $\alpha_{st}$  is the weighting parameter for surface t when the actual match is played on surface s

We do not propose to estimate these  $\alpha_{st}$  parameters in this paper.

#### 3. Markov Chain model

The basic principles involved in modelling a tennis match are well known, and a Markov chain model with a constant probability of winning a point was set up by Schutz (1970). While such a model is acceptable within a game, a model which allows a player a different probability of winning depending on whether they are serving or receiving is essential for tennis. Statistics of interest are usually the chance of each player winning, and the expected length of the match. Croucher (1986) looks at the conditional probabilities for either player winning a single game from any position. Pollard (1983) uses a more analytic approach to calculate the probability for either player winning a game or set along with the expected number of points or games to be played with their corresponding variance.

Most of the previous work uses analytical methods, and treats each scoring unit independently. This results in limited tables of statistics. Thus the chance of winning a game and the expected number of points remaining in the game is calculated at the various scores within a game. The chance of winning a set and the expected number of games remaining in the set is calculated only after a completed game and would not show for example the probability of a player's chance of winning from three games to two, 15-30.

This paper discusses the use of spreadsheets to repeat these applications using a set of interrelated spreadsheets. This allows any probabilities to be entered and the resultant statistics automatically calculated or tabulated. In addition, more complicated workbooks can be set up

which allow the calculation of the chance of winning a match at any stage of the match given by the point, game and set score. These allow the dynamic updating of player's chances as a match progresses.

Alternatively these algorithms could be converted into a programming language for automatic integration into multimedia as live scores are received.

#### **3.1 Game**

We explain the method by first looking at a single game where we have two players, A and B, and player A has a constant probability  $p_A$  of winning a point on serve. We set up a Markov chain model of a game where the state of the game is the current game score in points (thus 40-30 is 3-2). With probability  $p_A$  the state changes from a, b to a+1, b and with probability  $q_A=1-p_A$  it changes from a, b to a, b + 1. Thus if  $P_A^{pg}(a,b)$  is the probability that player A wins the game when the score is (a,b), we have:

$$\begin{split} &P_A{}^{pg}(a,b){=}p_AP_A(a{+}1,b){+}q_AP_A(a,b{+}1)\\ &\text{The boundary values are:}\\ &P_A{}^{pg}(a,b)=1 \text{ if } a=4,\,b\leq 2,\,P_A{}^{pg}(a,b)=0 \text{ if } b=4,\,a\leq 2. \end{split}$$

The boundary values and formula can be entered on a simple spreadsheet. The problem of deuce can be handled in two ways. Since deuce is logically equivalent to 30-30, a formula for this can be entered in the deuce cell. This creates a circular cell reference, but the iterative function of Excel can be turned on, and Excel will iterate to a solution. In preference, an explicit formula is obtained by recognizing that the chance of winning from deuce is in the form of a geometric series

$$\begin{split} P_{A}{}^{pg}(3,3) &= {p_{A}}^2 + {p_{A}}^2 2 p_A q_A + {p_{A}}^2 \left( 2 p_A q_A \right)^2 + {p_{A}}^2 \left( 2 p_A q_A \right)^3 + \dots \\ \text{where the first term is } p_{A}{}^2 \text{ and the common ratio is } 2 p_A q_A \end{split}$$

The sum is given by  $p_A^2/(1-2p_Aq_A)$  provided that  $-1<2p_Aq_A<1$ . We know that  $0<2p_Aq_A<1$ , since  $p_A>0$ ,  $q_A>0$  and  $1-2p_Aq_A=p_A^2+q_A^2>0$ .

Therefore the probability of winning from deuce is  $p_A^2/(1-2p_Aq_A)$ . Since  $p_A+q_A=1$ , this can be expressed as:

$$P_A^{pg}(3,3) = p_A^2/(p_A^2+q_A^2)$$

Excel spreadsheet code to obtain the conditional probabilities of player A winning a game on serve is as follows:

Enter  $\mathbf{p}_{\mathbf{A}}$  in cell D1

Enter  $\mathbf{q}_{\mathbf{A}}$  in cell D2

Enter 0.60 in cell E1

Enter =1-E1 in cell E2

Enter 1 in cells C11, D11 and E11

Enter **0** in cells G7, G8 and G9

Enter =  $E1^2/(E1^2+E2^2)$  in cell F10

Enter =\$**E**\$1\***C**8+\$**E**\$2\***D**7 in cell C7

Copy and Paste cell **C7** in cells D7, E7, F7, C8, D8, E8, F8, C9, D9, E9, F9, C10, D10 and E10

Notice the absolute and relative referencing used in the formula =\$E\$1\*C8+\$E\$2\*D7. By setting up an equation in this recursive format, the remaining conditional probabilities can easily and quickly be obtained by copying and pasting.

Table 2 represents the conditional probabilities of player A winning the game from various score lines for  $p_A = 0.60$ . It indicates that a player with a 60% chance of winning a point has a 74% chance of winning the game. Note that since advantage server is logically equivalent to 40-30, and advantage receiver is logically equivalent to 30-40, the required statistics can be found from these cells. Also worth noting is that the chances of winning from deuce and 30-30 are the same.

				B score		
		0	15	30	40	game
	0	0.74	0.58	0.37	0.15	0
	15	0.84	0.71	0.52	0.25	0
A score	30	0.93	0.85	0.69	0.42	0
	40	0.98	0.95	0.88	0.69	
	game	1	1	1		

Table 3: The conditional probabilities of A winning the game from various score lines

Similar equations can be developed for when player B is serving such that  $p_A$  and  $p_B$  represent constant probabilities of player A and player B winning a point on their respective serves. Also  $P_A^{pg}(a,b)$  and  $P_B^{pg}(a,b)$  represent the conditional probabilities of player A winning a game from point score (a,b) for player A and B serving in the game respectively.

A tennis match consists of four levels - (points, games, sets, match). Games can be standard games (as above) or tiebreak games, sets can be advantage or tiebreak, and matches can be the best-of-5 sets or the best-of-3 sets. To win a set a player needs six games with at least a two game lead. If the score reaches 6 games-all, then a tiebreak game is played in a tiebreak set to determine the winner of the set, otherwise standard games continue indefinitely until a player is two games ahead and wins the set. This latter scoring structure is known as an advantage set and is used as the deciding set in the Australian Open, French Open and Wimbledon. In some circumstances we may be referring to points in a standard or tiebreak game and other circumstances points in a tiebreak or advantage set. It becomes necessary to represent

points in a game as pg,

points in a tiebreak game as pg<sup>T</sup>,

points in an advantage set as ps,

points in a tiebreak set as ps<sup>T</sup> points in a best-of-5 set match (advantage fifth set) as pm, points in a best-of-5 set match (all tiebreak sets) as pm<sup>T</sup> games in an advantage set as gs, games in a tiebreak set as gs<sup>T</sup>, sets in a best-of-5 set match (advantage fifth set) as sm, and sets in a best-of-5 set match (all tiebreak sets) as sm<sup>T</sup>.

## 3.2 Tiebreak game

Since the chance of a player winning a tiebreak game depends on who is serving, two interconnected sheets are required, one for when player A is serving and one for when player B is serving. The equations that follow for modelling a tiebreak game, set and match are those for player A serving in the game. Similar formulas can be produced for player B serving in the game.

Let  $P_A^{pgT}(a,b)$  and  $P_B^{pgT}(a,b)$  represent the conditional probabilities of player A winning a tiebreak game from point score (a,b) for player A and player B serving in the game respectively.

Recurrence formulas:

$$\begin{split} &P_{A}{}^{pgT}(a,b) = p_{A}P_{B}{}^{pgT}(a+1,b) + q_{A}P_{B}{}^{pgT}(a,b+1), \text{ if } (a+b) \text{ is even} \\ &p_{A}{}^{pgT}(a,b) = p_{A}P_{A}{}^{pgT}(a+1,b) + q_{A}P_{A}{}^{pgT}(a,b+1), \text{ if } (a+b) \text{ is odd} \end{split}$$

Boundary values:

$$\begin{split} &P_{A}{}^{pgT}(a,b){=}1 \text{ if } a{=}7,\ 0{\le}b{\le}5 \\ &P_{A}{}^{pgT}(a,b){=}0 \text{ if } b{=}7,\ 0{\le}a{\le}5 \\ &P_{A}{}^{pgT}(6,6){=}p_{A}q_{B}/(p_{A}q_{B}{+}q_{A}p_{B}) \end{split}$$

where:  $q_B=1-p_B$ 

Table 4 represents the conditional probabilities of player A winning the tiebreak game from various score lines for  $p_A = 0.62$  and  $p_B = 0.60$ , and player A serving. Table 5 is represented similarly with player B serving.

Note how the calculations are obtained by the interconnection of both sheets. For example

$$\begin{split} P_A{}^{pgT}(0,0) &= p_A P_B{}^{pgT}(1,0) + q_A P_B{}^{pgT}(0,1) \\ &= 0.62 * 0.62 + 0.38 * 0.39 \\ &= 0.53 \end{split}$$

					B score				
		0	1	2	3	4	5	6	7
	0	0.53	0.44	0.29	0.20	0.10	0.04	0.01	0
	1	0.67	0.53	0.43	0.27	0.17	0.07	0.02	0
	2	0.76	0.68	0.53	0.42	0.24	0.13	0.03	0
A score	3	0.87	0.77	0.69	0.53	0.40	0.20	0.08	0
	4	0.93	0.89	0.80	0.72	0.52	0.37	0.13	0
	5	0.98	0.95	0.92	0.83	0.75	0.52	0.32	0
	6	0.99	0.99	0.98	0.96	0.89	0.82	0.52	
	7	1	1	1	1	1	1		

Table 4: The conditional probabilities of player A winning the tiebreak game from various score lines for  $p_A = 0.62$  and  $p_B = 0.60$ , and player A serving

					B score				
		0	1	2	3	4	5	6	7
	0	0.53	0.39	0.29	0.17	0.10	0.03	0.01	0
	1	0.62	0.53	0.37	0.27	0.14	0.07	0.01	0
	2	0.76	0.63	0.53	0.35	0.24	0.10	0.03	0
A score	3	0.83	0.77	0.63	0.53	0.33	0.20	0.05	0
	4	0.93	0.86	0.80	0.65	0.52	0.29	0.13	0
	5	0.97	0.95	0.89	0.83	0.67	0.52	0.21	0
	6	0.99	0.99	0.98	0.93	0.89	0.71	0.52	
	7	1	1	1	1	1	1		

Table 5: The conditional probabilities of player A winning the tiebreak game from various score lines for  $p_A = 0.62$  and  $p_B = 0.60$ , and player B serving

#### 3.3 Tiebreak set

Formulas are now given for a tiebreak set. Similar formulas can be obtained for an advantage set.

Let  $P_A^{gsT}(c,d)$  and  $P_B^{gsT}(c,d)$  represent the conditional probabilities of player A winning a tiebreak set from game score (c,d) for player A and player B serving in the game respectively.

### Recurrence formula:

$$P_A^{gsT}(c,d) = P_A^{pg}(0,0)P_B^{gsT}(c+1,d) + [1 - P_A^{pg}(0,0)]P_B^{gsT}(c,d+1)$$

# Boundary Values:

$$P_A^{gsT}(c,d)=1$$
 if c=6, 0\leq d\leq 4 or c=7, d=5  
 $P_A^{gsT}(c,d)=0$  if d=6, 0\leq c\leq 4 or c=5, d=7  
 $P_A^{gsT}(6,6)=P_A^{pgT}(0,0)$ 

Notice how the cell  $P_A^{pg}(0,0)$ , which represents the probability of winning a game, is used in the recurrence formula for a tiebreak set. Using the formulas given for a game and a tiebreak game conditional on the point score and a tiebreak set conditional on the game score, calculations are now obtained for a tiebreak set conditional on both the point and game score as follows.

Let  $P_A^{psT}(a,b:c,d)$  represent the probability of player A winning a tiebreak set from (c,d) in games, (a,b) in points and player A serving in the set. This can be calculated by:

$$\begin{split} &P_{A}{}^{psT}(a,b:c,d) = &P_{A}{}^{pg}(a,b)P_{B}{}^{gsT}(c+1,d) + [1-P_{A}{}^{pg}(a,b)]P_{B}{}^{gsT}(c,d+1), \text{ if } (c,d) \neq (6,6) \\ &P_{A}{}^{psT}(a,b:c,d) = &P_{A}{}^{pgT}(a,b), \text{ if } (c,d) = (6,6) \end{split}$$

#### 3.4 Match

Formulas are now given for a best-of-5 set match, where all sets are tiebreak sets. Similar formulation can be obtained for a best-5 set match, where the deciding fifth set is advantage. Formulation can also be obtained for a best-of-3 set match.

Let  $P^{smT}(e,f)$  represent the conditional probabilities of player A winning a best-of-5 set tiebreak match from set score (e,f).

Recurrence Formula:

$$P^{smT}(e,f) = P_A^{gsT}(0,0)P^{smT}(e+1,f) + [1-P_A^{gsT}(0,0)]P^{smT}(e,f+1)$$

**Boundary Values:** 

$$P^{\text{smT}}(e,f)=1 \text{ if } e=3,f\leq 2$$
  
 $P^{\text{smT}}(e,f)=0 \text{ if } f=3, e\leq 2$ 

Notice how the cell  $P_A^{gsT}(0,0)$ , which represents the probability of winning a tiebreak set, is used in the recurrence formula for a best-of-5 set match. Using the formulas given for a tiebreak set conditional on the point and game score and a best-of-5 set tiebreak match conditional on the set score, calculations are obtained for a best-of-5 set tiebreak match conditional on the point, game and set score as follows.

Let  $P_A^{pmT}(a,b:c,d:e,f)$  represent the probability of player A winning a tiebreak match from (e,f) in sets, (c,d) in games, (a,b) in points and player A serving in the match. This can be calculated by:

$${P_{A}}^{pmT}(a,\!b:\!c,\!d:\!e,\!f) \!\!=\!\! P_{A}{}^{psT}(a,\!b:\!c,\!d) P^{smT}(e+1,\!f) \! + \! [1 \! - \! P_{A}{}^{psT}(a,\!b:\!c,\!d)] P^{smT}(e,\!f+1)$$

Excel spreadsheet code was given directly in section 3.1 to obtain the conditional probabilities of player A winning a game on serve. Using the formulas given in section 3, spreadsheets can be

developed for a game with player B serving, tiebreak game, tiebreak set conditional on the game score and a best-of-5 set tiebreak match conditional on the set score. By assigning a value for p<sub>B</sub> to a cell (cell E3 for example), the probability of winning a match from the outset can be obtained for any probability value of p<sub>A</sub> and p<sub>B</sub> by changing the probability values given in cells E1 and E3. By adding additional formulas to the spreadsheet for a tiebreak set conditional on the point and game score (section 3.3) and for a best-of-5 set tiebreak match conditional on the point, game and set score (section 3.4), the chances of player's winning the set and match can be obtained conditional on who is currently serving, point score, game score and set score. An interactive tennis calculator to reflect this methodology available www.strategicgames.com.au.

### 4. Updating rule for serving statistic estimation

Whilst prior estimates of points won on serve may be reliable for the first few games or even the first set, it would be useful to update the prior estimates with what has actually occurred. We will use an updating system of the form where the proportion of initial serving statistics (X) is combined with actual serving statistics (Y) to give updated serving statistics (Z) at any point within the match.

$$Z = [ac / (ac+b)] X + [b / (ac+b)] Y$$
 (6)

where:

a represents the expected number of games remaining in the match b represents the number of games played c is a constant

Experimental results reveal that c = 2.5 is a suitable constant for best-of-5 and best-of-3 set matches. Note that the updating process occurs after each point. This method is outlined in Carlin and Louis (2000) in relation to Bayesian analysis and applied to tennis in Barnett (2006).

By assigning these values to a,b the following important properties are met:

- 1) More weighting on initial estimates towards the start of the match
- 2) The weighting increases for the actual statistics as the match progresses
- 3) The weighting towards the end of the match is asymptotic to the actual match statistics

Based on the methodology used in section 3 to obtain formulas for the chance of winning, the expected number of points remaining in a game, the expected number of games remaining in the set and the expected number of sets remaining in the match could be developed. This would allow calculations for the expected number of games remaining in the match as required in equation 6.

## 5. Sports Multimedia

The Markov Chain model outlined in section 3 along with the data analysis outlined in section 2 to determine the parameters for the model, is used in multimedia to obtain tennis predictions during a live match, and made available at <a href="http://sportsflash.com.au/">http://sportsflash.com.au/</a>. Two feature prediction products have been devised – Crystal Ball and Looking Glass. The Crystal Ball provides the chances of winning the match in progress in the form of a pie chart. The Looking Glass (similar format to a stock market chart) plots the chances of winning the match on a game-by-game basis (as in tennis) or every one-minute time interval (as in soccer). The graphical, visual and interactive properties of the Crystal Ball and Looking Glass could encourage spectators to engage with the predictions throughout a match in progress.

Figure 1 represents the predictions through the Looking Glass for the US Open men's final between Nadal and Djokovic. From the outset Djokovic had a 68.0% chance of winning the match. After Nadal won the 1<sup>st</sup> Set 6-4, the chances of Djokovic to win the match decreased to 36.7%. Djokovic won the 2<sup>nd</sup> Set 7-5 and the chances to win the match increased to 61.0%. This value is represented in figure 1 by using the interactive mouse-over feature, where solid dots are given for breaks of serve.

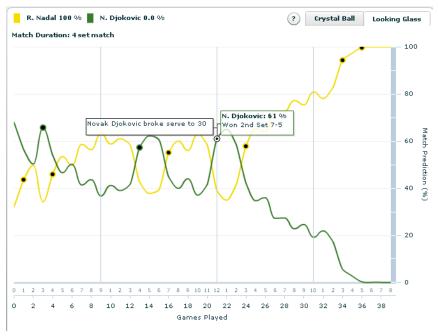


Figure 1: Live predictions for the US Open men's final between Rafael Nadal and Novak Djokovic

#### **Conclusions**

This paper demonstrates how spreadsheets can generate the probability of winning a tennis match conditional on the state of the match. These sheets have been used in multimedia to predict outcomes for a match in progress. There are many applications as to how these predictions could be used. An obvious application is in sports betting and the live predictions could provide a decision support tool to the punter. Another application is in using the predictions as a coaching tool for when players and coaches discuss strategies for upcoming matches on a replayed match. A further application is in using the live predictions for TV commentary by supporting the commentators' discussion on the likely winner of the match. The development of the predictions could also form an interesting and useful teaching example, and allow students to investigate the properties of tennis scoring systems.

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