

# Tiebreak proposals for Swiss tournaments

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Discussions about tiebreaks goes back to as far as 1873 (Ahrends, p. 184). Vivid discussions arose when the famous 1914 Mannheim chess tournament was interrupted by World War One. At that point Alekhine was a half point ahead. But Vidmar did have met stronger opponents. The issue of defining rankings for tournaments has been studied in various fields like sports, psychology, decision theory, Condorcet voting. Many ranking methods based on different motivations have been defined.

## Direct methods

### Solkoff, Buchholz, and variants

The above methods have a practical disadvantage. If a player leaves the tournament early the Solkoff contribution will be flatted in a negative sense. His opponents are thereby disadvantaged. The FIDE has found a solution for this (Forlani). But this solution is complicated and artificial. All direct methods have a fundamental limitation. They do take into account the direct opponents, but not the opponents of the opponents. And the opponents thereof. Four alternative indirect tiebreak methods are presented.

## Relative Elo Ratings

### The normal distribution function

"From general experience in sports we know that the stronger player does not invariably outperform the weaker. A player has good days and bad, good tournaments and bad. By and large at any point in his career, a player will perform around some average level. Deviations from this level occur, large deviations less frequently than small ones. These facts suggest the basic assumption of the Elo system. It is best stated in the formal terms of statistics:"

The many performances of an individual will be normally distributed, when evaluated on an appropriate scale (Elo, 1978, p. 19).

### Development of the Percentage Expectancy Table

The normal probabilities may be taken directly from the standard tables of the areas under the normal curve, when the difference  $D$  in ratings is expressed as a  $z$  score. The  $z$  value of a difference equals  $D / \sigma$ , where  $\sigma = 2000 / 7$ . For example, let  $D = 160$ . Then  $z = 160 * 7 / 2000 = 0,56$ . The table / formula gives ,7123 and ,2877 as areas of the two portions under the cumulative curve. These numbers represents the expected scores of the two opponents (Elo, p. 158).

In a rating system (KNDB, FMJD) ratings are updated at regular time intervals. Relevant games are considered. Game by game the difference between actual and expected score is established, according to the percentage expectancy function. Then the existing rating is updated, proportional to the total difference between expected and actual score. However, when computing the relative Elo ratings, we do not assume any existing ratings prior to the tournament. All players in the tournament are assigned a rating, the relative Elo rating, such that for each player the expected score and the actual score are equal. This rating is calculated by a numeric iteration procedure, as no closed formula exists. The relative Elo rating of a tournament is unique, up to a constant.

## Example [Open de Sangmelima Cameroon 2014](#)

At the start of the iterations (R0) one assumes all players have equal strength. In the subsequent iterations the rating of each player is updated according to  $-\ln(\text{We}/W) * 2000 / 7$ . "We" is the expected score derived from the cumulative normal distribution, W is the actual score and  $\ln()$  constitutes the natural logarithm function. The updates are normalized to set the sum of all updates to zero. The iteration is slow, but stable. We may choose at start any average rating level including zero. This does not affect the ordering of the players.

Pl	Title	Name	Cn	Rating	+	=	-	N	Pts	Whl	Wl		R0	We0	Up0		R1	We1	Up1		R2	We2	Up2		Rk	R28	
1	gmi	Leopold Kouomou	cm	2296	3	3	0	6	9	25	34		0,0	6,0	137,7		137,7	8,0	32,6		170,3	8,1	36,2		2	304,0	
2		Gerard Ngankou	cm	2265	3	3	0	6	9	25	34		0,0	6,0	137,7		137,7	7,7	42,9		180,6	8,0	38,3		1	312,7	
3		Mouanji Iliassou	cm		4	0	2	6	8	16	23		0,0	6,0	104,0		104,0	9,0	-33,9		70,1	8,5	-10,6		6	18,3	
4		Armand Abouem	cm		3	1	2	6	7	30	39		0,0	6,0	65,9		65,9	6,0	41,7		107,6	6,5	25,6		3	185,3	
5		Landry Nga	cm		1	5	0	6	7	28	37		0,0	6,0	65,9		65,9	6,3	31,1		97,0	6,6	20,5		4	180,9	
6	mi	Bruno Fopa	cm	2262	2	3	1	6	7	21	28		0,0	6,0	65,9		65,9	7,7	-28,3		37,5	7,5	-12,8		9	-26,6	
7		Desire Ghuendou	cm		2	3	1	6	7	18	25		0,0	6,0	65,9		65,9	8,2	-45,6		20,3	7,6	-17,6		10	-63,0	
8		Maturin Tomi	be		2	2	2	6	6	26	35		0,0	6,0	21,8		21,8	6,4	-17,1		4,7	6,0	6,0		5	33,6	
9		Arnaud Foto	cm		2	2	2	6	6	25	32		0,0	6,0	21,8		21,8	6,3	-15,9		6,0	6,2	-3,6		8	-19,4	
10		Patrick Akono	cm		1	3	2	6	5	31	40		0,0	6,0	-30,3		-30,3	4,4	36,2		5,9	5,2	-5,6		7	-16,3	
11		Bernard Mambo	cm		2	0	4	6	4	28	37		0,0	6,0	-94,0		-94,0	3,7	18,8		-75,3	4,2	-8,6		11	102,5	
12		David Wabo Daco	cm		1	2	3	6	4	23	31		0,0	6,0	-94,0		-94,0	4,7	-45,7		-139,7	4,1	-3,3		12	177,2	
13		Barel Bimogo	cm		1	1	4	6	3	23	31		0,0	6,0	-176,2		-176,2	3,8	-66,9		-243,1	3,1	-5,4		13	299,4	
14		Prince Mvondo	cm		1	0	5	6	2	28	37		0,0	6,0	-292,1		-292,1	1,7	50,2		-241,8	2,5	-59,1		14	330,6	

It is easy to check the results of the above calculation:

Gerard Ngankou	313,0			Leopold Kouogueu Kouomou	304,0		
Opponents	R28	Diff	We	Opponents	R28	Diff	We
Patrick Akono	-16,3	329,3	1,75	Prince Arnaud Mvondo	-330,6	634,6	1,97
Landry Nga	180,9	132,1	1,36	Maturin Nyamsi Tomi	33,6	270,4	1,66
Maturin Nyamsi Tomi	33,6	279,4	1,67	Patrick Akono	-16,3	320,3	1,74
Bernard Mambo	-102,5	415,5	1,85	Armand Abouem	185,3	118,7	1,32
Leopold Kouogueu Kouomou	304,0	9,0	1,03	Gerard Ngankou	312,7	-8,7	0,98
Armand Abouem	185,3	127,7	1,35 +	Landry Nga	180,9	123,1	1,33 +
Expected Score			9,0 points	Expected score			9,0 points

## Game file

Pl	Name	Rtg	R1	R2	R3	R4	R5	R6	Pt
1	gmi Leopold Kouogueu Kouomou	2296	2/14b	1/8b	2/10w	2/4w	1/2b	1/5b	9
2	Gerard Ngankou	2265	1/10b	1/5b	2/8w	2/11b	1/1w	2/4b	9
3	Mouanji Iliassou		0/4w	0/10b	2/13b	2/14b	2/12w	2/11b	8
4	Armand Abouem		2/3b	2/9b	2/11w	0/1b	1/5w	0/2w	7
5	Landry Nga		1/8w	1/2w	1/12b	2/9w	1/4b	1/1w	7
6	mi Bruno Fopa	2262	0/12w	1/7b	2/14w	2/8b	1/10b	1/9w	7
7	Desire Ghuendou		0/11w	1/6w	1/9b	1/13b	2/14w	2/10b	7
8	Tomi Maturin Nyamsi		1/5b	1/1w	0/2b	0/6w	2/13w	2/14b	6
9	Arnaud Foto		2/13b	0/4w	1/7w	0/5b	2/11w	1/6b	6
10	Patrick Akono		1/2w	2/3w	0/1b	1/12b	1/6w	0/7w	5
11	Bernard Mambo		2/7b	2/12w	0/4b	0/2w	0/9b	0/3w	4
12	David Daco Wabo		2/6b	0/11b	1/5w	1/10w	0/3b	0/13w	4
13	Barel Bimogo		0/9w	0/14b	0/3w	1/7w	0/8b	2/12b	3
14	Prince Arnaud Mvondo		0/1w	2/13w	0/6b	0/3w	0/7b	0/8w	2

[FMJD report](#)

## Domain

The iteration requires an indivisible (that is, strongly connected or irreducible) domain. In every possible partition of players into two nonempty subsets, some player of the second set has defeated at least one player in the first set (Glickman, p. 5). Or alternatively, there exists a directed path between any two players. Finding all maximal strongly connected components is a mathematical puzzle in its own right.

Remark: one can introduce a virtual opponent who draws to all other players to make the domain indivisible.

## Skew-symmetric score systems

A score system  $R$  is skew-symmetric if for all possible outcomes  $(x, y)$   $x + y = 0$ . Any traditional score system can be transformed to this form (Chebotarev, 1989, p. 1104):

$$\text{Let } A = \begin{vmatrix} |x & 3 & .| \\ |0 & x & 1| \\ |. & 1 & x| \end{vmatrix}, \text{ transpose}(A) = \begin{vmatrix} |x & 0 & .| \\ |3 & x & 1| \\ |. & 1 & x| \end{vmatrix}, \text{ then } R = \begin{vmatrix} |x & 1\frac{1}{2} & .| \\ |-1\frac{1}{2} & x & 0| \\ |. & 0 & x| \end{vmatrix}$$

where

- $A$  represents the football score; win point = 3.
- $R = (A - \text{transpose}(A)) / 2$ .
- Divide by two to preserve the difference between own and opponent score.

# Least squares method

## Principle

If a player has a win, loss or draw his comparison outcome is set to 1, -1 or 0 respectively. The least squares method constructs a mean square approximation of the skew-symmetric comparison outcomes  $R$  by the differences of the desired least squares ratings (Cheboratev, 1999, p. 18):

Determine  $q = (q_1, \dots, q_n)$  such that  $\sum_i \sum_j (r_{ij} - (q_i - q_j)) n_{ij}^2$  is a minimum (Gulliksen, 1956)

where

- $n_{ij}$  is the number of games between player  $i$  and  $j$ ,  $i \neq j$ , zero otherwise (matches matrix).
- $r_{ij}$  equals the sum of the results of player  $i$  against  $j$ .

## Calculation

The LSM-ratings ( $q$ ) are a solution of the set of linear equations:

- $L \cdot q = s$ .
- $\sum q = 0$ .

where

- $L = (l_{ij})$  is the Laplacian matrix of the tournament graph.
- $l_{ii} =$  Number of comparisons of player  $i$ .
- $l_{ij} =$  Negative the number of games between player  $i$  and  $j$ ,  $i \neq j$ .
- $s = (s_i)$ ,  $s_i = \sum_k r_{ik}$ .
- Skew-symmetric draughts scores  $r_{ik}$  are equivalent to (wins -/- losses).

## Round-robin tournament

LSM-ratings  $q$  of a round-robin tournament:

- $q = s / (N + r)$ , where  $r = 1, 2, 3$  for a single, double, triple round-robin tournament.

**Note** that if the percentage expectancy function of the relative Elo ratings is replaced by a linear relationship, for example the "algorithm of 400", then relative Elo ratings are identical to LS-ratings except for a scaling factor. Multiplying LS-ratings with a factor  $(400 / R_{\max})$  gives a comparable rating in the Elo domain (FIDE Rating Regulations effective from 1 July 2017, table 8.1b).

## Example Open de Sangmelima Cameroon 2014

Pl	Title	Name	Cn	Rating	+ N	s	Whl	Wl	Rk	LS-Rtg	
1	gmi	Leopold Kouomou	cm	2296	3	6	3	25	34	2	0,643
2		Gerard Ngankou	cm	2265	3	6	3	25	34	1	0,708
3		Mouanji Iliassou	cm		4	6	2	16	23	5	0,059
4		Armand Abouem	cm		3	6	1	30	39	3	0,420
5		Landry Nga	cm		1	6	1	28	37	4	0,392
6	mi	Bruno Fopa	cm	2262	2	6	1	21	28	9	-0,042
7		Desire Ghuendou	cm		2	6	1	18	25	11	-0,122
8		Maturin Tomi	be		2	6	0	26	35	6	0,054
9		Arnaud Foto	cm		2	6	0	25	32	10	-0,047
10		Patrick Akono	cm		1	6	-1	31	40	8	-0,030
11		Bernard Mambo	cm		2	6	-2	28	37	12	-0,234
12		David Wabo Daco	cm		1	6	-2	23	31	13	-0,425
13		Barel Bimogo	cm		1	6	-3	23	31	15	-0,694
14		Prince Mvondo	cm		1	6	-4	28	37	14	-0,683

It is easy to check the result of the above calculation as follows:

Gerard Ngankou	0,708	3
Opponents	LS-Rtg	Diff
Patrick Akono	-0,030	0,738
Landry Nga	0,392	0,316
Maturin Nyamsi Tomi	0,054	0,654
Bernard Mambo	-0,234	0,942
Leopold Kouogueu Kouomou	0,643	0,065
Armand Abouem	0,420	0,288
s1	3,00	(Wins minus losses)

Leopold Kouogueu Kouomou	0,643	3
Opponents	LS-Rtg	Diff
Prince Arnaud Mvondo	-0,683	1,326
Maturin Nyamsi Tomi	0,054	0,589
Patrick Akono	-0,030	0,673
Armand Abouem	0,420	0,223
Gerard Ngankou	0,708	-0,065
Landry Nga	0,392	0,251 +
s2	3,00	(Wins minus losses)

# Recursive Buchholz

Let

- Rb - Recursive Buchholz ratings
- RbOpp - Average opponents recursive Buchholz ratings
- s - ( $s_i$ ) is sum of the skew-symmetric score points  $r_{ik}$
- p - Average of s

Rb, RbOpp, s, p are  $n \times 1$  column vectors indexed by player. The recursive Buchholz rating Rb is the result of the following iteration procedure: (González-Díaz et al., 2013, p. 6).

- $Rb = RbOpp + p$ .
- $\Sigma Rb = 0$ .

## Example Open de Sangmelima Cameroun 2014

Pl	Title	Name	Cn	Rating	+	=	-	N	Pts	Whl	Wl		[,1]	[,2]	[,3]	[,4]	[,5]	[,6]	[,7]	[,8]	[,9]	[,10]	[,11]	[,12]	...	[,16]--
1	gmi	Leopold Kouomou	cm	2296	3	3	0	6	9	25	34		0	0.5000	0.5000	0.6250	0.6103	0.6416	0.6358	0.6436	0.6417	0.6436	0.6431	0.6436	...	0.6435
2		Gerard Ngankou	cm	2265	3	3	0	6	9	25	34		0	0.5000	0.5556	0.6759	0.6728	0.7036	0.6993	0.7075	0.7056	0.7079	0.7071	0.7078	...	0.7077
3		Mouanji Iliassou	cm		4	0	2	6	8	16	23		0	0.3333	0.0278	0.1157	0.0478	0.0729	0.0553	0.0626	0.0578	0.0600	0.0586	0.0593	...	0.0590
4		Armand Abouem	cm		3	1	2	6	7	30	39		0	0.1667	0.3611	0.3611	0.4105	0.4070	0.4188	0.4173	0.4201	0.4196	0.4203	0.4201	...	0.4202
5		Landry Nga	cm		1	5	0	6	7	28	37		0	0.1667	0.3056	0.3380	0.3781	0.3791	0.3907	0.3889	0.3924	0.3914	0.3925	0.3921	...	0.3923
6	mi	Bruno Fopa	cm	2262	2	3	1	6	7	21	28		0	0.1667	0.0000	0.0046	-0.0324	-0.0328	-0.0397	-0.0404	-0.0414	-0.0419	-0.0418	-0.0421	...	0.0421
7		Desire Ghuendou	cm		2	3	1	6	7	18	25		0	0.1667	-0.0833	-0.0509	-0.1173	-0.1048	-0.1222	-0.1174	-0.1223	-0.1206	-0.1220	-0.1214	...	0.1217
8		Maturin Tomi	be		2	2	2	6	6	26	35		0	0.0000	0.0278	0.0370	0.0486	0.0507	0.0526	0.0536	0.0535	0.0540	0.0538	0.0540	...	0.0540
9		Arnaud Foto	cm		2	2	2	6	6	25	32		0	0.0000	-0.0278	-0.0324	-0.0455	-0.0419	-0.0474	-0.0450	-0.0471	-0.0460	-0.0468	-0.0464	...	0.0466
10		Patrick Akono	cm		1	3	2	6	5	31	40		0	-0.1667	0.0556	-0.0648	-0.0046	-0.0401	-0.0227	-0.0329	-0.0278	-0.0307	-0.0292	-0.0300	...	0.0298
11		Bernard Mambo	cm		2	0	4	6	4	28	37		0	-0.3333	-0.1944	-0.2593	-0.2215	-0.2423	-0.2300	-0.2370	-0.2329	-0.2352	-0.2339	-0.2347	...	0.2344
12		David Wabo Daco	cm		1	2	3	6	4	23	31		0	-0.3333	-0.3889	-0.3981	-0.4221	-0.4169	-0.4256	-0.4224	-0.4255	-0.4240	-0.4251	-0.4245	...	0.4248
13		Barel Bimogo	cm		1	1	4	6	3	23	31		0	-0.5000	-0.5833	-0.6667	-0.6690	-0.6907	-0.6876	-0.6941	-0.6922	-0.6943	-0.6934	-0.6941	...	0.6940
14		Prince Mvondo	cm		1	0	5	6	2	28	37		0	-0.6667	-0.5556	-0.6852	-0.6559	-0.6853	-0.6772	-0.6843	-0.6820	-0.6838	-0.6832	-0.6836	...	0.6835

Note:

- The procedure converges iff the matches matrix is not bipartite (Csató, 2015, p. 8).
- Although the perspective of RB and LSM are quite different, both ratings coincide (González-Díaz et al., 2013, p. 10).

# Generalized row sum method

## Principle

Row sum scores ( $s$ ) are calculated as the sum over  $k$  of the skew-symmetric outcomes  $r_{ik}$ . The generalized row sum method is an extension of the row sum method. The idea of the generalization is to complete the missing comparisons. Central to the statistical interpretation of the generalized row sum method is the assumption that the expected value  $E(r_{ij})$  of a missing comparison is proportional to the difference of the scores  $x_i, x_j$  of the compared players  $i$  and  $j$ :

- $x_i = s_i + \sum_j E(r_{ij})$ , where  $r_{ij}$  is missing,  $s_i = \sum_k r_{ik}$  where  $r_{ik}$  is defined and
- $E(r_{ij}) = (x_j - x_i) / \gamma$ , for any positive factor  $\gamma$ , see Chebotarev, 1989, Chapter 3 Statistical model, p. 1105.

These relationships form a system of  $n$  linear equations in  $n$  unknowns  $x_i$ . For  $\gamma = m \cdot n + \varepsilon^{-1}$  the system of equations is equivalent to:

- $x_i = \sum_k (r_{ik} + \varepsilon(x_k - x_i + r_{ik} \cdot n \cdot m))$ ,  $k = 1, \dots, n$ .

where

- $\varepsilon$  is a positive parameter.
- $\varepsilon$  is well-chosen for given  $n$  and  $m$  if and only if  $\varepsilon^{-1} \geq m(n - 2)$ .
- $r_{ik} + r_{ki} = 0$ , the skew-symmetric outcome between players  $i, k$ .
- number of players  $n > 2$ ,  $m =$  maximal number of games between any two players.

**Monotonicity.** Parameter  $\varepsilon$  is said to be well-chosen if for any outcome matrix  $R = (r_{ik})$  the value of its contribution to  $x_i$  is nonnegative for a maximal win and nonpositive for a maximal loss (Chebotarev, 1997, Ch. 5, p. 9).

## Calculation

The generalized row sums ( $x$ ) are a solution of the set of linear equations:

- $(L + \varepsilon^{-1}I)x = \gamma s$ .
- $\sum x = 0$ .

where

- $L = (l_{ij})$  is the Laplacian matrix of the matches matrix,  $I$  is the identity matrix.
- $\gamma = m \cdot n + \varepsilon^{-1}$ .

Note that LS-ratings can be obtained as a limit of the generalized row sum calculation by setting  $\varepsilon^{-1} = 0$  and  $\gamma = 1$ .

## Example Trainingsvierkamp Harm Wiersma Huizum 2005 (Blitz)

n = 4						m = 2, 1/ε = 2×4											
Pl	Naam	Rating	1	2	3	4	N	Matches	L + (1/ε)I	+	=	-	Pts	s	Rk	x	
1	Harm Wiersma	GMI 1502	xx	02	22	..	4	0 2 2 0	12 -2 -2	0	3 0 1	6	2	1	2,667		
2	Rein van der Pal	MF 1402	20	xx	02	12	6	2 0 2 2	-2 14 -2 -2	3	1 2	7	1	2	1,000		
3	Tjalling van den Bosch	1190	00	20	xx	21	6	2 2 0 2	-2 -2 14 -2	2	1 3	5	-1	3	-1,000		
4	Jan Adema	1217	..	10	01	xx	4	0 2 2 0	0 -2 -2 12	0	2 2	2	-2	4	-2,667		

x = column vector of generalized row sum ratings  
s = column vector wins -/- losses

### Statistical model

Check expected values  $x_i = s_i + \frac{\sum(x_i - x_j)}{\gamma}$ ,  $\gamma = 1/\epsilon + n.m$ ,  $\gamma = 16$   
All players are expected to complete an equal number of games

Harm Wiersma	2,67	Jan Adema	-2,67
s1	2	s4	-2
Missing games		Missing games	
(x1-x4)/γ	0,333	(x4-x1)/γ	-0,333
(x1-x4)/γ	0,333	(x4-x1)/γ	-0,333
	----- +		----- +
s1 + missing gms	2,67	s1 + missing gms	-2,67

### Direct method

Check  $(L + (1/\epsilon)I)x = \gamma s$ ,  $1/\epsilon = 8$ ,  $\gamma = 16$

	L+(1/ε)I	x	(L + (1/ε)I)x
Harm Wiersma	12	2,6667	32,0004
Rein van der Pal	-2	1,0000	-2,0000
Tjalling van den Bosch	-2	-1,0000	2,0000
Jan Adema	0	-2,6667	0,0000
		----- +	
γ.s1			32,0000
s1			2,000

- The statistical model is useful when the tournament is nearly complete.
- For sparse tournament matrices, the direct method is more practical.



## Self-consistency (SC) and self-consistent monotonicity (SCM)

*Self-consistency* comes down to the following. Suppose we consider two alternatives (players), having the same number of comparisons, and the first alternative as compared to the second one

- achieves the better scores against the “stronger” opponents or
- achieves the better scores against the opponents of the same “strengths” or
- achieves the same scores against the “stronger” opponents.

Then the first alternative should be placed higher than the second one in the tiebreak ranking. In the above requirement, “stronger” signifies” is “placed higher in the ranking that is mentioned in the requirement”.

To obtain *self-consistent monotonicity* we require that if the first alternative additionally achieves some extra “wins” and/or the second alternative has extra “losses”, then the first alternative should remain higher than the second one in the tiebreak ranking. (Chebotarev, 1997, p. 2)

Direct methods as row sum score, Buchholz, Sonneborn-Berger (Neustadt!) are monotonic. Direct methods are rather indifferent to opponent strength and therefore incompatible with self-consistency. Recursive Buchholz and LSM satisfy self-consistency but contradicts monotonicity. Relative Elo ratings and generalized row sum ratings obey self-consistent monotonicity.

Consider the "algorithm of 400", the AVG400 rating system. If the rating difference between player and opponent exceeds 400, the expected plus score percentage is greater the 50%. However this is impossible to achieve. Therefore a strong player can lose points even with a perfect score and a weak player can gain by losing all his games (Elo, p. 144).

## References

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