



Electronic Monitoring and Assessment Resources for Coaching: Digital empowerment of sport coaches

# **Observational and** Notational Analysis in Sports

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Brazil



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**Ipvc** sprint nstituto Politécnico de Viana do Castelo Recreation, Innovation and Technology







Escuela Municipal de Deportes Villanueva del









UFMG Graduated in Physical Education UNIVERSIDADE FEDER



UF<sup>m</sup>G Master in Sports Sciences



UF<sup>*m*G</sup> Ph.D in Sports Sciences



**CBF** Coach Educator

### **FIFA** FIFA Coach Educator



Head of the CECA/UFMG



U F *M* G

UNIVERSIDADE FEDERAL DE MINAS GERAIS





# Agenda

• Basic concepts on Observational Analysis



- Paying attention to the details!
- Practical example of coding technical-tactical events
- Research on observational analysis in sports

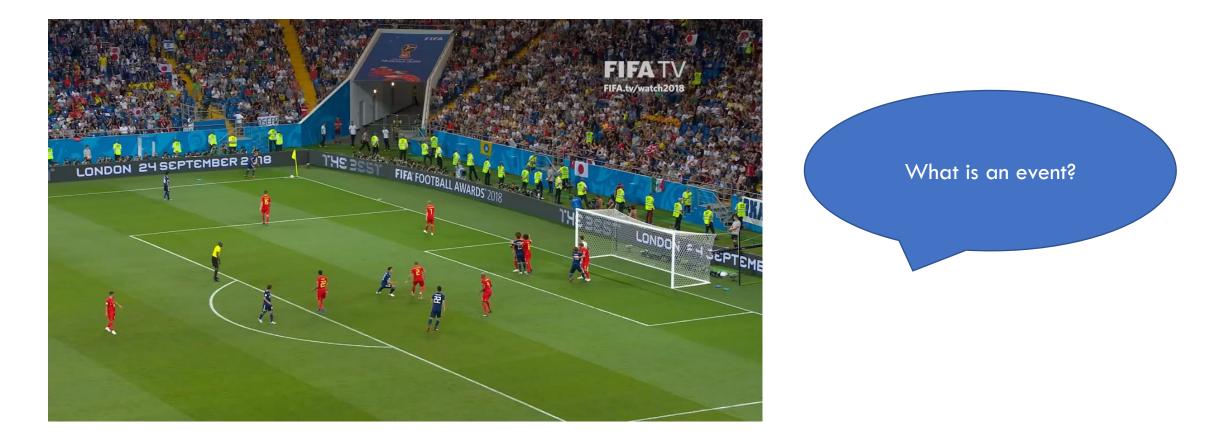


UNIVERSIDADE FEDERA







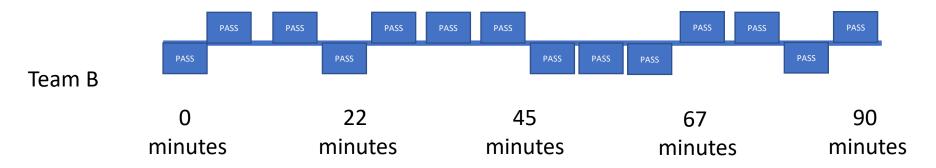






#### Team A

CECA







 Notational analysis is a means of recording events in which there is an objective and precise process to describe what actually happened. In this context, notational analysis allows a reliable record that "does not lie" (of course, if the process of collecting information is reliable and if the observation system allows answering the right questions) (Carling, Williams and Reilly, 2008).

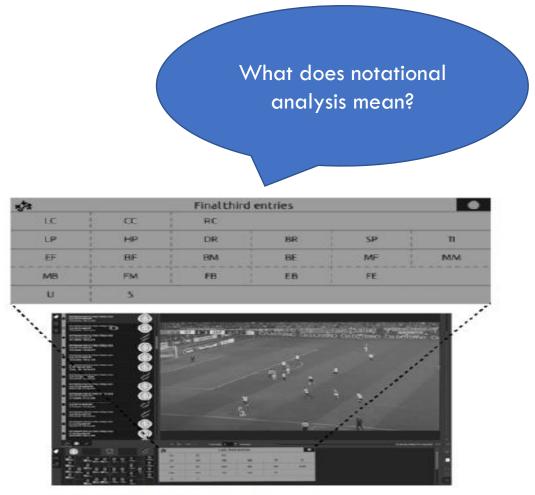


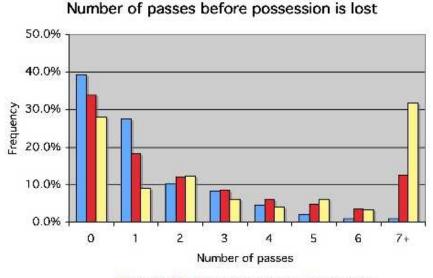
Figure 1. Dashboard developed in the LongoMatch Software.









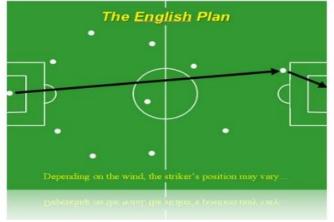


<sup>🔲 1953-67 📕</sup> Prem league 2012 🛄 Spain u21s

#### The distant past?



end in less than 4 passes



The importance of interpreting the data contextually.

<u>Source</u>





### Definitions of tactical and technical-related events

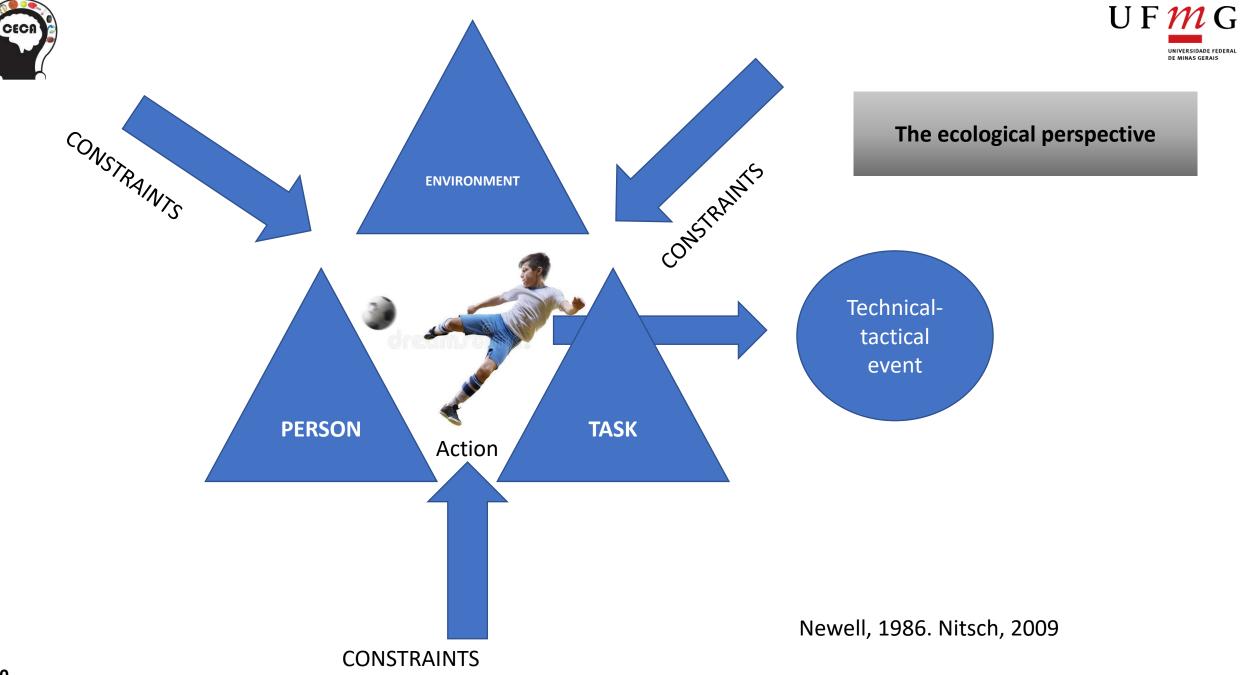
The traditional perspective



Tactical event: the observable decision-making action.



Technical event: the observable motor action.







Ensuring internal consistency periodically





			Absolute	Difference in
	Moment 1	Moment 2	difference	%
Event A	2	3	1	50%
Event B	2	2	0	0%
Event C	4	5	1	25%
Event D	6	8	2	33%
Event E	7	7	0	0%
Event F	9	6	3	33%
		Average Difference	1,167	24%

Internal consistency: the consistency between observations made by the same observer, in different moments





Ensuring external consistency periodically





			Absolute	Difference in
	Observer 1	Observer 2	difference	%
Event A	2	3	1	50%
Event B	2	2	C	0%
Event C	4	5	1	25%
Event D	6	8	2	33%
Event E	7	7	C	0%
Event F	9	6	3	33%
		Average Difference	1,167	24%

External consistency: the consistency between observations made by two different observers during the same events.



Gibson Moreira Praça, Cristóvão de Oliveira Abreu, Marcelo Rochael and Pedro Drumond Moreira

Variable	ICC	95% confidence interval Lower bound	Upper bound	SEM	SEM%	p-value
Length (m)	0.841	0.711	0.918	1.865	9.77	< 0.001
Width (m)	0.911	0.839	0.954	1.368	8.68	< 0.001*
SEL (m)	0.500	0.063	0.752	0.833	11.92	0.015*
FUT-SAT Offensive	0.355	-0.159	0.664	16.15	23.41	0.071
FUT-SAT Defensive	0.119	-0.583	0.541	15.41	25.37	0.329
GPET Decision-making	-0.944	-2.495	-0.014	0.189	27.96	0.977
GPET Execution	-0.172	-1.107	0.389	0.176	27.29	0.680
SNA Degree Centrality	0.204	-0.431	0.585	0.083	40.89	0.219
SNA Degree Prestige	0.423	-0.037	0.699	0.083	36.09	0.033*
SNA Page Rank	0.361	-0.150	0.667	0.212	29.12	0.067

Variable	ICC	95% confidence interval Lower bound	Upper bound	SEM	SEM%	þ-value	
Length (m)	0.933	0.808	0.977	0.800	4.19	< 0.001*	
Width (m)	0.944	0.545	0.985	0.888	5.63	< 0.001*	
SEL (m)	0.705	0.156	0.897	0.467	6.68	0.012*	
FUT-SAT Offensive	0.755	0.293	0.915	6.18	8.96	0.002*	
FUT-SAT Defensive	0.307	-0.481	0.725	8.53	14.04	0.188	
GPET Decision-making	0.415	-0.674	0.796	0.054	7.99	0.155	
GPET Execution	0.651	0.062	0.875	0.054	8.37	0.011*	
SNA Degree Centrality	0.307	-0.907	0.758	0.044	21.67	0.243	
SNA Degree Prestige	0.506	-0.414	0.827	0.054	23.48	0.092	
SNA Page Rank	0.387	-0.755	0.786	0.130	17.86	0.177	

SEI: Spatial exploration index; SNA: Social Network Analysis. \*Significant differences (p < 0.05)

Table 4. Between-sessions reliability values.

#### SEI: spatial exploration index; SNA: Social Network Analysis.

\*Significant differences (p < 0.05)





### Reporting the most adequate summary measures

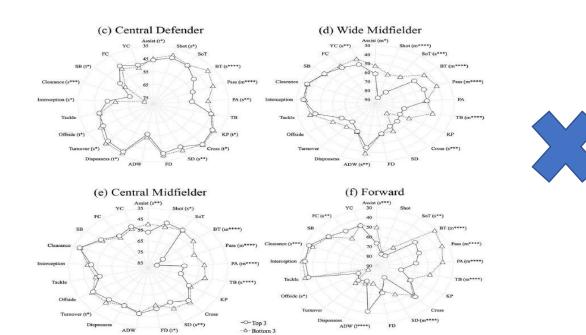


Table 1. Descriptive statistics of match performance profiles of players from Top3 and Bottom3 teams (results are counts, except for Pass Accuracy).

	All Players $(n = 1583)$		Full Back ( $n = 382$ )		Central Defender $(n = 415)$		Wide Midfielder $(n = 183)$		Central Midfielder $(n = 402)$		Forward ( $n = 201$ )	
Variable	Top3	Bottom3	Top3	Bottom3	Top3	Bottom3	Top3	Bottom3	Top3	Bottom3	Top3	Bottom3
Assist	0.2 ± 0.4	0.1 ± 0.2	0.1 ± 0.3	0.0 ± 0.2	0.0 ± 0.2	0.0 ± 0.1	0.3 ± 0.5	0.1 ± 0.3	0.2 ± 0.4	0.1 ± 0.3	0.3 ± 0.6	0.1 ± 0.4
Shot	1.4 ± 1.9	$1.1 \pm 1.4$	0.6 ± 0.9	$0.4 \pm 0.7$	$0.6 \pm 0.8$	$0.4 \pm 0.7$	4.0 ± 3.2	2.0 ± 1.4	0.9 ± 1.1	1.2 ± 1.3	3.3 ± 2.0	3.1 ± 2.0
Shot on target	0.5 ± 1.0	$0.4 \pm 0.7$	$0.2 \pm 0.4$	0.1 ± 0.3	$0.2 \pm 0.4$	$0.2 \pm 0.4$	1.5 ± 1.6	0.7 ± 0.8	$0.3 \pm 0.6$	0.3 ± 0.6	1.5 ± 1.3	1.1 ± 1.0
Ball touch	71 ± 27	53 ± 16	79 ± 24	59 ± 15	58 ± 21	48 ± 15	63 ± 18	53 ± 13	86 ± 30	59 ± 18	61 ± 24	41 ± 10
Passes	52 ± 26	34 ± 15	50 ± 22	30 ± 12	44 ± 21	32 ± 14	41 ± 15	31 ± 10	70 ± 30	45 ± 18	44 ± 23	26 ± 8
Pass Accuracy (%)	82 ± 11	75 ± 12	83 ± 10	73 ± 12	81 ± 14	75 ± 13	78 ± 9	78 ± 10	85 ± 9	77 ± 11	82 ± 9	70 ± 11
Through ball	$0.5 \pm 1.0$	$0.1 \pm 0.4$	0.2 ± 0.5	0.0 ± 0.1	$0.1 \pm 0.3$	$0.1 \pm 0.3$	0.9 ± 1.1	0.1 ± 0.4	0.8 ± 1.2	$0.2 \pm 0.5$	0.9 ± 1.2	0.2 ± 0.5
Key pass	$1.0 \pm 1.3$	0.9 ± 1.2	0.9 ± 1.0	0.6 ± 0.8	0.2 ± 0.5	$0.2 \pm 0.4$	$2.1 \pm 2.0$	1.6 ± 1.4	1.3 ± 1.3	1.2 ± 1.4	1.4 ± 1.1	1.4 ± 1.6
Crosses	$1.8 \pm 2.6$	2.1 ± 3.2	2.9 ± 2.7	$2.5 \pm 2.3$	0.1 ± 0.4	0.1 ± 0.3	4.5 ± 4.4	6.3 ± 4.4	1.8 ± 2.4	1.6 ± 2.9	1.3 ± 1.5	1.7 ± 2.5
Successful dribble	0.9 ± 1.4	$0.6 \pm 0.9$	0.8 ± 1.1	$0.4 \pm 0.8$	0.3 ± 0.5	0.1 ± 0.4	1.5 ± 1.6	1.2 ± 1.2	0.9 ± 1.2	0.6 ± 0.9	$2.0 \pm 2.0$	1.0 ± 1.2
Foul drawn	$1.2 \pm 1.3$	$1.2 \pm 1.3$	1.2 ± 1.1	0.8 ± 1.0	0.6 ± 0.9	0.5 ± 0.8	1.9 ± 1.6	1.9 ± 1.5	1.3 ± 1.2	1.5 ± 1.3	1.9 ± 1.7	2.1 ± 1.4
Aerial duel Won	1.5 ± 1.7	1.7 ± 1.8	1.2 ± 1.4	1.3 ± 1.2	2.2 ± 1.9	2.0 ± 1.6	1.2 ± 1.7	0.7 ± 1.0	1.4 ± 1.6	1.5 ± 1.8	0.9 ± 1.3	3.5 ± 2.7
Dispossessed	1.1 ± 1.3	0.9 ± 1.2	1.0 ± 1.0	0.5 ± 0.7	0.2 ± 0.5	0.2 ± 0.4	1.6 ± 1.4	1.6 ± 1.3	1.0 ± 1.2	1.1 ± 1.1	2.3 ± 1.7	2.2 ± 1.4
Turnover	0.9 ± 1.2	0.7 ± 1.0	0.8 ± 0.9	0.6 ± 0.7	0.3 ± 0.5	0.2 ± 0.5	1.8 ± 1.5	1.4 ± 1.2	0.9 ± 1.0	0.7 ± 0.8	1.9 ± 1.6	1.7 ± 1.5
Offside	$0.3 \pm 0.7$	$0.2 \pm 0.6$	0.1 ± 0.4	0.1 ± 0.3	$0.1 \pm 0.3$	$0.0 \pm 0.2$	0.6 ± 1.0	0.4 ± 0.7	$0.1 \pm 0.4$	0.1 ± 0.3	0.9 ± 1.1	1.2 ± 1.2
Tackle	$2.1 \pm 1.8$	2.1 ± 1.7	2.7 ± 2.0	2.8 ± 1.8	2.0 ± 1.5	$2.0 \pm 1.6$	1.7 ± 1.7	1.3 ± 1.2	2.7 ± 2.1	$2.5 \pm 1.9$	0.9 ± 1.0	0.6 ± 0.9
Interception	$1.7 \pm 1.6$	$1.7 \pm 1.7$	2.2 ± 1.7	2.1 ± 1.7	2.1 ± 1.5	2.5 ± 1.9	$0.6 \pm 0.8$	0.8 ± 1.0	1.9 ± 1.7	1.7 ± 1.5	$0.5 \pm 0.8$	0.3 ± 0.6
Clearance	2.7 ± 3.3	3.7 ± 3.9	2.5 ± 2.0	4.0 ± 2.6	6.3 ± 3.7	8.0 ± 4.2	0.6 ± 1.0	0.7 ± 0.9	1.5 ± 1.8	1.5 ± 1.7	0.3 ± 0.8	0.9 ± 1.5
Shot block	0.3 ± 0.6	0.3 ± 0.6	0.3 ± 0.5	0.3 ± 0.5	0.5 ± 0.8	0.6 ± 0.9	0.1 ± 0.2	0.1 ± 0.3	0.3 ± 0.6	0.3 ± 0.6	0.1 ± 0.3	0.1 ± 0.3
Foul committed	1.3 ± 1.3	1.3 ± 1.3	1.3 ± 1.4	1.2 ± 1.1	1.0 ± 1.1	1.2 ± 1.1	1.1 ± 1.2	1.0 ± 1.0	1.6 ± 1.4	1.7 ± 1.6	1.2 ± 1.3	1.7 ± 1.5
Yellow card	0.2 ± 0.4	0.3 ± 0.4	0.1 ± 0.4	0.3 ± 0.5	0.2 ± 0.4	0.3 ± 0.4	0.2 ± 0.4	0.1 ± 0.3	0.3 ± 0.4	0.3 ± 0.5	0.2 ± 0.4	$0.2 \pm 0.4$











Creating online backups and dashboards







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Live Transmission Settings		н	AC Milan (4.)	Udinese Calcio (17						90'			Duván Zapata	=	20
Account Settings		A 	AC Milan (4.) AC Milan (3.)	Sampdorla (9.) Inter Mediolan (4.)						90'			Cristiano Ronaldo		19
Log Out		н	AC Milan (3.)	inter Mediolan (4.						90'			Arkadiusz Milik	-	16







**OBSERVATIONAL METHODOLOGY IN SPORT SCIENCES** 

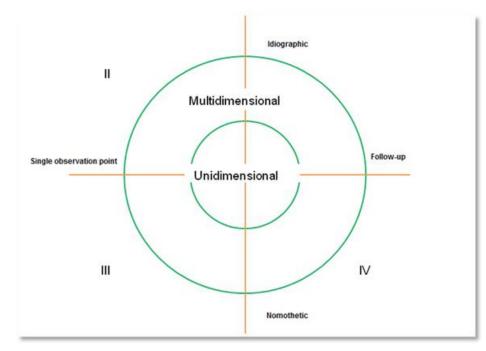
La metodología observacional en el ámbito del deporte

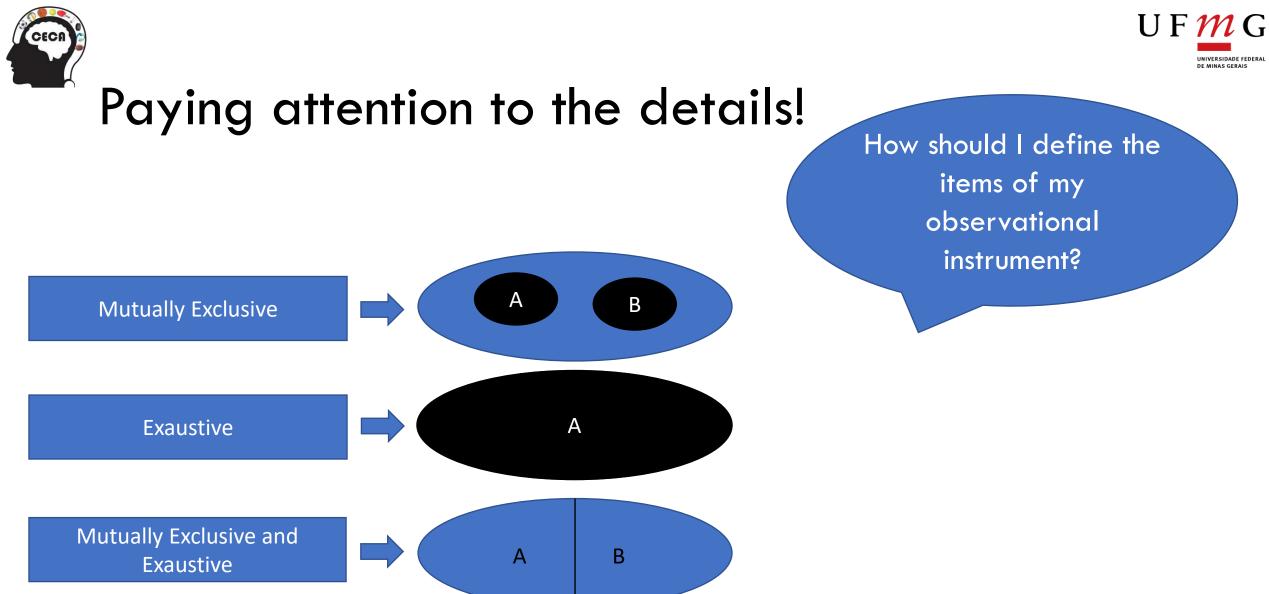
M. Teresa Anguera University of Barcelona

Antonio Hernández Mendo University of Málaga

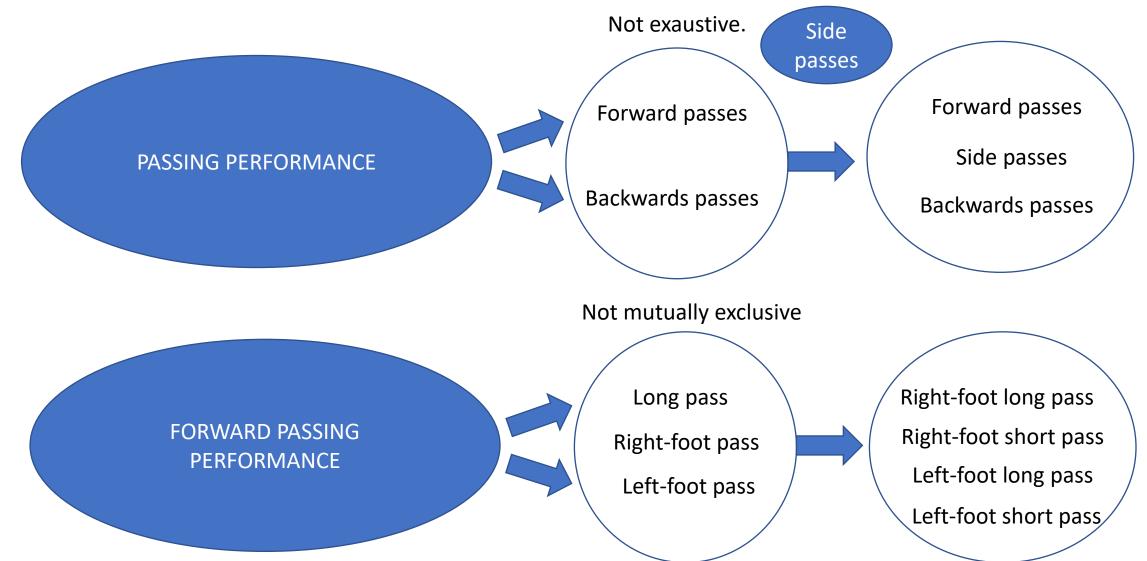
Correspondence:

Antonio Hernández Mendo Facultad de Psicología. Campus de Tectinas, s/n. Universidad de Málaga. 29071 MÁLAGA (SPAIN) Mail: mendo@uma.es Recibido: 10/08/2013 Aceptado: 16/10/2103 How should I define the items of my observational instrument?













### Practical example of coding technical-tactical events





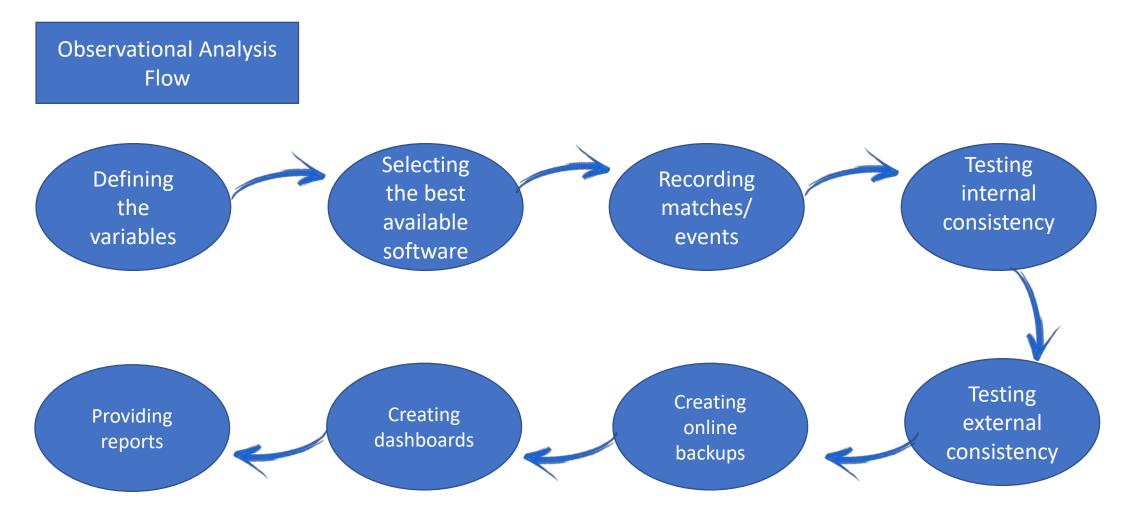
LONGOMATCH https://longomatch.com/en/pro/

LINCE PLUS https://observesport.github.io/lince-plus/





### Practical example of coding technical-tactical events









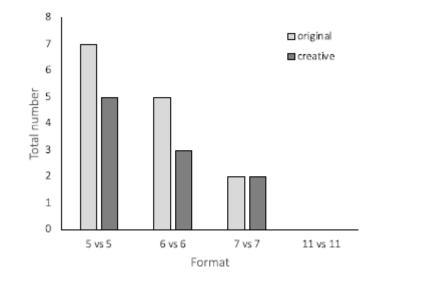
Psychology of Sport and Exercise Volume 48, May 2020, 101645



Technical events

Variability and creativity in small-sided conditioned games among elite soccer players \*

Simone Caso, John van der Kamp 🙁 🖾



Small-sided games with fewer players induce a higher frequency of creative on-the-ball actions. No creative actions were observed in the formal 11-a-side match.

The observational data support using smaller SSG formats to nurture creativity in soccer players.





**Technical events** 

### Research on observational analysis in sports

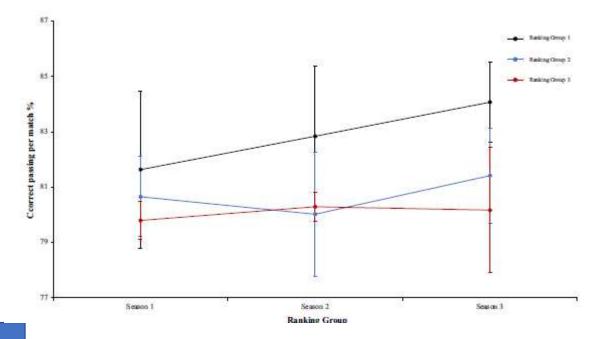
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### scientific reports

### OPEN Variations in the physical demands and technical performance of professional soccer teams over three consecutive seasons

Zeki Akyildiz<sup>1</sup>, Hadi Nobari<sup>2,303</sup>, Francisco Tomás González-Fernández<sup>4</sup>, Gibson Moreira Praça<sup>5</sup>, Hugo Sarmento<sup>6</sup>, Aytek Hikmet Guler<sup>7</sup>, Esat Kaan Saka<sup>8</sup>, Filipe Manuel Clemente<sup>9,10</sup> & António J. Figueiredo<sup>11</sup>



Elite teams are passing the ball more and more successfully over the seasons.



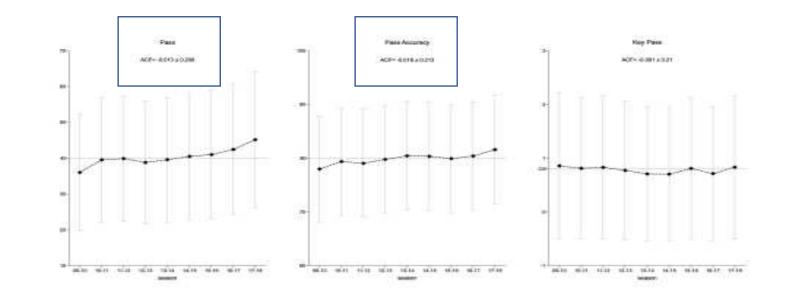


- update

### Evolutionary Trends of Players' Technical Characteristics in the UEFA Champions League

Qing Yi1,2,3\*, Hongyou Liu₄, George P. Nassis1,5 and Miguel-Ángel Gómez<sup>€</sup>

<sup>1</sup> School of Physical Education and Sport Training, Shanghai University of Sport, Shanghai, China, <sup>2</sup> Shanghai Key Lab of Human Performance, Shanghai University of Sport, Shanghai, China, <sup>3</sup> Key Laboratory of Diagnosis & Analysis of Skills & Tactics in Sports, Shanghai University of Sport, Shanghai, China, <sup>4</sup> School of Physical Education & Sports Science, South China Normal University, Guangzhou, China, <sup>9</sup> Department of Sports Science and Clinical Biomechanics, Faculty of Health Sciences, SDU Sport and Health Sciences Cluster, University of Southarn Denmark, Odense, Denmark, <sup>9</sup> Facultad de Clencias de la Actividad Fisica y del Deporte (INEF), Universidad Politécnica de Madrid, Madrid, Spain Players are performing more passes and showing a higher pass accuracy over time.





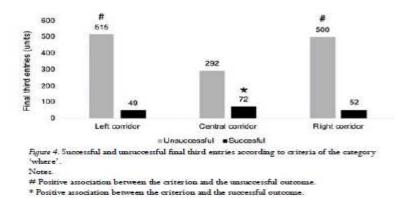


2022, Retos, 43, 171-176 © Copyright:Federación Española de Asociaciones de Doomtes de Educación Física (FEADEF)

ISSN:Edición impresa:1579-1726.EdiciónWeb:1988-2041 (https://recyt.fecyt.es/index.php/retos/index)

### How Reaching the Pitch's Final Third is Related to Scoring Opportunities in Soccer?

Cómo la invasión del último tercio del campo se relaciona con las oportunidades de marcar en el fútbol? João Pedro Araújo Guimarães, Marcelo Rochael, André Gustavo Pereira de Andrade, Sarah da Glória Teles Bredt, Gibson Moreira Praça Universidade Federal de Minas Gerais (Brazil)



There are fewer final third entries in the central corridor, although they are more successful there than in the lateral corridors.

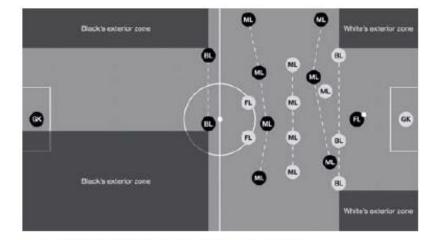


Figure 2. Illustration of spatial pattern of interaction between teams. Notes. GK: goalkeeper; BL: back line; MF: midline; FM: offensive line.

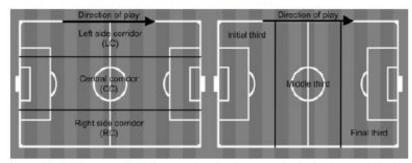


Figure 3. Pitch corridors and thirds.







HUMAN MOVEMENT (ISSN 1899-1955) 2021; 22(2): 22–31

#### THE INFLUENCE OF PLAYING POSITION ON THE PHYSICAL, TECHNICAL, AND NETWORK VARIABLES OF SUB-ELITE PROFESSIONAL SOCCER ATHLETES

original paper DOI: https://doi.org/10.5114/hm.2020.100010

### GIBSON PRAÇA<sup>1</sup>, LAURA BEATRIZ FALEIRO DINIZ<sup>1</sup>, FILIPE MANUEL CLEMENTE<sup>2,3</sup>, SARAH DA GLÓRIA TELES BREDT<sup>1</sup>, BRUNO COUTO<sup>1,4</sup>,

© University School of Physical Education in Wroclaw

ANDRÉ GUSTAVO PEREIRA DE ANDRADE<sup>1</sup>, ADAM LEE OWEN<sup>5,6</sup>

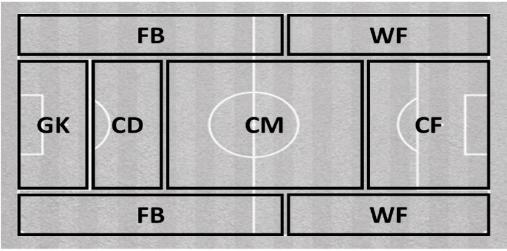
<sup>1</sup> Universidade Federal de Minas Gerais, Belo Horizonte, Brazil

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<sup>3</sup> Instituto de Telecomunicações, Delegação da Covilhã, Covilhã, Portugal

<sup>4</sup> University of the Sunshine Coast, Queensland, Australia <sup>5</sup> Claude Bernard Lyon University, Lyon, France

<sup>6</sup> Claude Bernard Lyon University, Lyon, France
<sup>6</sup> Seattle Sounders Football Club, Seattle, Washington, USA



GK - goalkeeper, CD - central defender, FB - fullback, CM - central midfielder, WF - wide forward, CF - central forward

Table 2. Players' network measures and technical responses depending on playing position

Measure	Central defenders	Fullbacks	Central midfielders	Wide forwards	Central forwards	р	Effect size	Pairwise comparisons
Degree centrality (AU)	0.16 (0.04)	0.14 (0.05)	0.07 (0.03)	0.03 (0.01)	0.02 (0.01)	0.001*	0.651	CD, FB > CM, WF, CF
Degree prestige (AU)	0.06 (0.03)	0.10 (0.03)	0.10 (0.04)	0.13 (0.05)	0.09 (0.04)	0.023*	0.242	CD < WF
Page rank (AU)	0.04 (0.03)	0.08 (0.02)	0.11 (0.04)	0.12 (0.01)	0.15 (0.03)	0.001*	0.509	CD < CM, WF, CF; FB < CF
Successful passes (n)	41.58 (22.38)	43.10 (17.37)	25.85 (15.68)	15.00 (5.47)	9.17 (3.81)	0.001*	0.383	CD, FB > CF
Successful passes (%)	83.37 (19.83)	79.19 (7.70)	75.44 (12.63)	70.23 (12.61)	63.21 (15.63)	0.085	0.181	
Shots (n)	0.67 (0.65)	0.56 (0.72)	2.54 (2.02)	1.75 (0.50)	3.17 (1.47)	0.001*	0.400	CD, FB < CM, CF
Lost balls (n)	0.25 (0.45)	2.33 (1.80)	3.23 (1.83)	4.50 (1.29)	2.83 (2.13)	0.001*	0.462	CD < FB, CM, WF, CF
Recovered balls (n)	2.42 (1.83)	3.80 (1.39)	4.00 (2.51)	3.25 (2.21)	1.33 (1.03)	0.047	0.209	

AU – arbitrary units, CD – central defender, FB – fullback, CM – central midfielder, WF – wide forward, CF – central forward \* significant differences

The closer the opposing goal, the less successful the passes are.





INTERNATIONAL JOURNAL OF PERFORMANCE ANALYSIS IN SPORT https://doi.org/10.1080/24748668.2020.1753979 Routledge Taylor & Francis Group

Check for update:

### Players' match demands according to age and playing position in professional male soccer players

Changjing Zhou 00<sup>e</sup>, Alberto Lorenzo 00<sup>e</sup>, Miguel-Ángel Gómez 00<sup>e</sup> and José M. Palao<sup>b</sup>

\*Facultad de Ciencias de la Actividad Física y del Deporte (INEF), Universidad Politécnica de Madrid, Madrid, Spain; \*Department of Health, Exercise Science and Sport Management, University of Wisconsin-Parkside, Kenosha, WI, USA Players from those highly physicallydemanded positions experience a more prominent age-related drop in performance.

Table 1. Data distribution according to age group and playing position (values expressed in occurrence and percentage).

Position	Group 1 (17-22.9 yr)	Group 2 (23-28.9 yr)	Group 3 (29-34.9 yr)	Group 4 (35-40.9 yr)	Total
Central defender	516 (9%)	2,929 (49%)	2,211 (37%)	350 (6%)	6,006
Fullback	659 (12%)	3,591 (63%)	1,337 (24%)	85 (1%)	5,672
Central midfielder	702 (10%)	3,517 (48%)	3,022 (41%)	148 (2%)	7,389
Wide midfielder	897 (16%)	3,476 (63%)	1,062 (19%)	72 (196)	5,507
Attacker	262 (7%)	2,045 (51%)	1,579 (40%)	87 (2%)	3,973
Total	3,036 (11%)	15,558 (54%)	9,211 (32%)	742 (3%)	28,547

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#### Technical and physical match performance of teams in the 2018 FIFA World Cup: Effects of two different playing styles

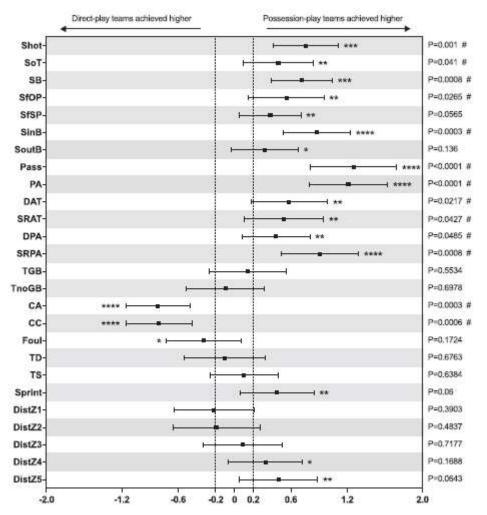
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> Possession-play teams achieve more shots, shots on target, passes, and passing accuracy than direct-play teams in the World Cup.

Access the article here

#### Direct-play teams - Possession play teams





# Take Home Message



Ensuring the quality of your observational data is mandatory.

Getting high-quality videos will facilitate the data-gathering process.

Dealing with tagging software requires practice. Take your time!

More than getting lots of information, asking the right questions will improve the quality of the analysis.







"I've never scored a goal in my life without getting a pass from someone else" (Abby Wambach)

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