1	Interactive effects of emotions on performance: An exploratory study in elite skeet
2	shooters <sup>1</sup>
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4 5 6 7 8 9 10 11 12 13 14	INTERACTIVE EFFECTS OF EMOTIONS ON PERFORMANCE: AN EXPLORATORY STUDY IN ELITE SKEET SHOOTERS KEYWORDS: IZOF model, performance prediction, emotion interaction, aggregated scale. ABSTRACT: This study explored interactive effects of multiple emotions on performance (outcomes) in three male elite skeet shooters. Recalled emotional experiences associated to successful and poor performances and to current (actual) performances were measured using the aggregated Emotional State Profile-40 scale (ESP-40; Hanin, 2010b). Current performance involved 20 series of 25 shots. The results provide support for the practical utility of ESP-40 in the assessment of interactive effects of four emotion categories on performance. Recalled measures were instrumental in the prediction of current performances. The utility of ordinal logistic regression-based estimation is discussed.
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16 17	Author Note This study is the result of collabora concetween Ruiz & Hanin (as supervisors) and Fogaça & Pääk (former master's students). The data presented in the article were
18 19 20 21 22 23 24 25 26 27 28 29 30	This study is the result of collaboration between Ruiz & Hanin (as supervisors) and Fogaça & Pääk (former master's students). The data presented in the article were collected (by the team coach) using an unpublished version of a questionnaire developed by Hanin (ESP-40). Hanin and Kuiz provide the questionnaire and assessment protocol to the students who instructed the loach in administration procedures. The entire "sample" that the students present in their masters thesis included three shooters, from whom emotion and affect data were collected. The article, however, presents an in-depth analysis of emotion data from one shooter used as an illustration of emotion dynamics. Cooperation with the students was not possible. There was mutual oral agreement on data publication. Written permission from students, however, was not materialized. Such requirement was not normal tractice of the university. This practice is now being introduced.

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An accurate description of athletes' emotional experiences associated with individual
 success and failure is important for the development of effective intervention programs
 (Hanin, 2000, 2007, 2010a; Robazza, Bortoli, Nocini, Moser, & Arslan, 2000; Ruiz &
 Hanin, 2004b).

5 Previous research focused on the study of single emotions such as competitive 6 anxiety (Jones, 1995; Martens, Vealey, & Burton, 1990; Raglin & Hanin, 2000). 7 Cognitive and somatic components of anxiety were believed to impact performance differently. Hard (1990, 1996) assumed that while physiological arousal and low 8 cognitive anxiety followed an inverted-U relationship with performance, as cognitive 9 10 anxiety increased, performance could improve up to a critical point, after which it suddenly declined. Although this as umption implies interaction effects, these 11 components were assessed separately. Other studies indicate that anger associated to 12 successful performances was perceived as pen incial or detrimental for athletes 13 14 depending on the intensity and the situational meaning (Ruiz & Hanin, 2004a, 2011). Thus, a need to examine pleasant and unpleasant emotions in the prediction of athletic 15 performance was highlighted (Hanin, 2000, 2004, 2007, 2015a; Robazza, Bortoli, & 16 17 Hanin, 2004). Methodologically, research has focused on the assessment of single emotions 18

disregarding their interactive effects. For example, existing standardized scales include
the Competitive State Anxiety Inventory-2 (CSAI-2; Martens, Burton, Vealey, Bump,
& Smith, 1990), the Sport Emotion Questionnaire (SEQ, Jones, Lane, Bray, Uphill, and
Catlin, 2005) or the Profile of Mood States (POMS; McNair, Lorr, & Droppleman,

23 1971). Although the so called "iceberg profile" characterized by high values for vigor

1	and low values for tension, confusion, depression, anger and fatigue may imply
2	interactive effects, the athletes' rate the intensity for each item separately.
3	Research also revealed a discrepancy between the content of items in
4	standardized scales, and the vocabulary used by athletes. For example, 80-85% of
5	content of individual emotional experiences of soccer players was not assessed by
6	standardized instruments (Hanin, 2007; Syrjä, 2000). A study with elite karate athletes
7	indicated individual preferences in selection of idiosyncratic labels to describe
8	performance-related anger states (Ruiz & Hanin 2004a).
9	In other word, group-oriented scales are limited in capturing the athlete's
10	perspective or personal merainy. Thus, an individualized approach termed the
11	Individual Zones of Optimax Functioning (IZOF) model (Hanin, 1997, 2000, 2004,
12	2007, 2010a) was advocated in the sudv of athletes' emotions. The IZOF model
13	distinguishes emotional states (experiences per se), relatively stable emotional patterns
14	(repeated experiences), and meta-experiences (i.e., inowledge, beliefs, and attitudes
15	about recalled, actual or anticipated experiences) (Hania, 2004, 2007). Emotion is
16	conceptualized as a situational, multi-modal and dynamic munifestation of the total
17	human functioning (Hanin, 1997, 2000). The content of emotions is conceptualized
18	within the framework of two related factors: functioning (success-facure) and hedonic
19	tone or valence (pleasure-displeasure) resulting in success-related functionally optimal
20	pleasant (P+) and unpleasant (N+) emotions and failure-related dysfunctional
21	unpleasant (N-) and pleasant (P-) emotions. These four categories serve to identify
22	emotion labels relevant for performance reflecting readiness to perform (Hanin, 1997,
23	2000, 2010a).

1	An example of sport-specific individualized emotion measures is individualized
2	emotion profiling (IEP, Hanin, 2000). In IEP, athletes identify individual and task-
3	relevant content and intensity of emotional experiences (pleasant and unpleasant)
4	accompanying successful and poor performances. With the help of a stimulus list,
5	athletes generate idiosyncratic emotion descriptors based on four emotion categories.
6	Idiographic profiles represent interactive effects of optimal and dysfunctional emotions
7	where optimal emotions are placed in the middle and dysfunctional emotions by their
8	sides. An "iceberg" emotional profile, typical in successful performances, is
9	characterized by higher intensities of functionally optimal emotions. In contrast, a
10	"cavity" emotional profile cypical of poor performances, is characterized by higher
11	intensities of dysfunctional emotion.
12	A more recent approach involves aggregating (most often selected) self-
13	generated labels across athletes and sport evens and the four emotion categories (N- N+
14	P+P-). Emotion State Profile (ESP-40, Hanin, 2010) consists of an aggregated 40 item
15	scale with ten items in each of the four emotion categories to assess interactive effects
16	between these emotion categories. Thus, ESP-40 captures id osynstatic content better
17	than standardized psychometric scales. In addition, emotion items are similar for all
18	athletes making possible between-individual comparisons across similar categories and
19	across similar items within each emotion category. Similar to IEP, functional optimal
20	categories of ESP-40 items are placed in the middle while dysfunctional categories are
21	on the sides to facilitate a visual representation of interactive effects. Optimal emotional
22	profiles related to success are iceberg (bell-shaped), and may indicate predominance of
23	functionally pleasant emotions represented by the sequence N- $<$ N+ $<$ P+ $>$ P-, or
24	functionally unpleasant emotions (N- $<$ N+ $>$ P+ $>$ P-). In contrast, dysfunctional

1 emotion profiles related to failure are flat or skewed profiles which reflect

 $2 \qquad \mbox{predominance of unpleasant dysfunctional emotions (N->N+>P+>P-) \ \mbox{or pleasant}} \\$ 

3 dysfunctional emotions (N- < N+ < P+ < P-).

4 Within the IZOF model, performance predictions are based on the "in-out of the 5 zone" notion (Hanin, 2000). High probability of successful performance is expected 6 when individual emotion content and intensity fall within optimal and outside 7 dysfunctional zones previously established. Thus, predominance of optimal emotions 8 and low levels or absence of dysfunctional emotions is expected in successful performances while performance of dysfunctional emotions and absence of optimal 9 10 emotions is expected in unsuccessful performances. Empirical support for these assumptions was obtained in creas-country skiers (Hanin, 1997), gymnasts (Cottyn, De 11 Clercq, Crombez, & Lenoir, 2012 12 occur and ice hockey players (Hagtvet & Hanin, 2007), swimmers and track & field athletes (Robezza, Pellizari, Bertollo, & Hanin, 13 14 2008). Thus, the purpose of this investigation was to examine the interactive effects of 15 emotions related to athletic performance in three elite skeet mooters. Specifically, this 16 investigation examined the practical utility of an aggregated emotion scale, the ESP-40 17 scale, in the assessment of four emotion categories and their interactive effects on 18

19 performance. In addition, this study explored the practical utility of recalled emotional

- 20 experiences before successful and poor performances in the prediction of performance.
- 21

#### Method

## 22 **Participants**

23 Participants were three male skeet shooters. This was a convenience sample, with

24 criterion for inclusion being that participants were experienced and represented the

1	highest level o	of achievement in	their modality. Partic	pipants, with ages ranging	ig from 18
2	to 27 years, we	ere members of th	e Finnish National T	eam. They had from 3 t	o 13 years
	of internationa	al experience, and	had achieved medals	in major international	
	competitions (	e.g. European Ch	ampionships, World	Championships, or Wor	ld Cups),
	being among t	he most successfu	Il athletes in their eve	ent.	
	Measures				
	The <i>Emotion</i>	State Profile (ESI	P-40; Hanin, 2010b, s	see appendix, Fogaça &	Päkk,
	2012) is a 40-i	iten scale that ass	esses functionally op	timal pleasant (P+), fun	ctionally
	optimal unplea	asant (N+), dysfur	nctional pleasant (P-)	, and dysfunctional unp	leasant (N-)
	emotions. ESF	P-40 consists of a	list of aggregated em	otion labels drawn from	most often
	selected words	s describing atalet	s' gates before or d	uring athletic performar	nce.
	Specifically, E	ESP-40 consists of	10 rows of 4 column	ns with one adjective for	r each
	emotion categ	ory. To capture th	e interactive effects,	a within row compariso	n is used.
	Participants ra	nk each item base	ed on how accurated	it describes their emotion	onal state
	from 4 (descri	bes best) to 1 (des	cribes least). Scores	of + and 3 represent the	presence of
	emotion where	eas scores 1 and 2	represent low intens	ity or absence of emotic	on. An
	example of rai	nking is provided	below:		
	N-	N+	P+	Р-	
	[1] Tired	[3] Tense	[4] Energetic	[2] Easy-going	
				es in each column (emo (maximum), are visuall	

1	represented in an emotion profile, with functional optimal categories (N+, P+) placed in
2	the middle and dysfunctional categories (N-, P-) by the sides to facilitate a visual
3	representation of interactive effects. Optimal profiles have an iceberg form (or bell-
4	shape), whereas dysfunctional profiles are flat or skewed. The ranking order in each row
5	is an important indicator of specific interaction pattern across the emotion categories.
6	This interaction can be also represented in a rank of emotion categories distinguishing
7	four emotion profiles: N- <n+<p+>P- (type one – optimal positive), N-<n+>P+&gt;P-</n+></n+<p+>
8	(type two – optimal negative), N->N+>P+>P- (type three – dysfunctional negatively
9	skewed), and N- $N+q+P-$ (type four – dysfunctional positively skewed) <sup>2</sup> .
10	Performance measures. Shooting performance consists of series of 25 shots. Scores (0
11	= missing the target; $1 = hit$ , recorded at the end of a series of 25 targets, range from 0
12	to 25.
13	Procedure
14	The participants, recruited by the coach, gave written informed consent in accordance
15	with APA ethical guidelines. Due to the study characteristics and nature of the sport,
16	data were collected by the coach who was instructed in the use of the measures and
17	procedures. Data collection took place during three training camps that lasted from 6 to
18	7 days each, organized within three months before the competition scason. Before the
19	first training camp, the shooters were asked to recall three most successful performances
20	and to describe how they felt before each using the ESP-40. Then, they did the same for

<sup>&</sup>lt;sup>2</sup> Previous research (Bortoli, Bertollo, Hanin, & Robazza, 2012) has indicated that types one and two are usually related to optimal performances, while types three and four are typically related to poor performances.

using the ESP-40, 30 min before 20 series of 25 shots. Shooting scores were recorded
after each series.

#### 3 Data Analysis

4 Individual emotion profiles were developed for successful, poor and current 5 performances. Shooting scores were categorized as better than standard performance 6 (scores 24 - 25), standard performance (scores 22 - 23), and sub-standard performance 7 (scores 21 and below). In performance prediction, current (actual) emotions were 8 contrasted against previously recalled emotions associated to successful and unsuccessful perform aces. Optimal zones of functioning were determined using min -9 10 max ranges of emotion intersities for each category before three most successful competitions, and three most up accessful competitions for dysfunctional zones. 11 Distances between emotion intensity levels before actual performance and previously 12 established optimal and dysfunctional zones with calculated. Probabilistic estimation 13 approach (see Kamata, Tenenbaum & Hanin, 2002) applies ordinal logistic regression 14 (OLR) models, where performance outcomes are assured of categorical nature (e.g. 15 optimal, non-optimal), and emotional intensity is the predictor variable. This method 16 assumes the probability that non-optimal performance is associated with emotion 17 intensities above or below intensity levels in optimal current performance. Logistic 18 19 curves represent the relationships between probabilities and performance outcomes. 20 Finally, Spearman's rank correlation coefficients were calculated for inter-correlations 21 between 10 combinations based on four emotion categories: N-, N+, P+, P-, pleasant 22 (P++P-), unpleasant (N++N-), helpful (P++N+), harmful (P-+N-), strong 23 unpleasant (N+ - N-), and strong pleasant (P+ - P-). An inter-correlation matrix was 24 used to draw maximum correlational paths (Vyhandu, 1964).

1	Results
2	Due to space limitations main results for one shooter will be presented. Additional data
3	for the other two shooters were similar and will be summarized where appropriate.
4	Individual emotion profiles and data (for shooter A) before most successful and
5	unsuccessful performances are depicted in Figure 1. As expected, before successful
6	competitions, A reported predominance of pleasant (optimal and dysfunctional)
7	emotions. Characterized by the following emotion interactions N- <n+<p+>P- (8 out of</n+<p+>
8	9 possible patterne). In contrast, unsuccessful performances were characterized by N-
9	≤N+>P+ <p- emotion="" interactions.<="" th=""></p->
10	A's actual performance were better than standard on 11 occasions (55% of 20
11	series), standard performances (20%) and sub-standard (5%). Median values and min-
12	max ranges for emotions before A's current performances are presented in Table 1.
13	Emotional profiles for this shooter reflected N <n+<p+>P- (type one – optimal</n+<p+>
14	positive) interactions before better than standard and standard performances. However,
15	before below standard performance emotion interactions were N- <n+<p+<p 2<="" figure="" th=""></n+<p+<p>
16	presents boxplots of emotion intensities before three levels of current performance
17	compared with recalled measures. Interestingly, very low variability for emotion
18	intensities was experienced. In addition, emotion intensities before recalled successful
19	performances (represented by bars) were very close to those experienced before current
20	acceptable standard and better than standard performances.
21	Figure 3 illustrates probability curves for optimal performance based on emotion
22	intensities before current performances. As it can be seen, highest probabilities of
23	optimal performance were associated with absence or low intensities of emotions

24 whereas predominance of pleasant emotions predicted 70% probability of optimal

1 performance. Table 2 presents lower and upper values for performance prediction using

2 two estimation methods. Interestingly, an overlap was observed for success-related

3 zones for N- and N+ emotions. However, for P+ and P- emotions, optimal ranges fell

4 outside those actually experienced.

## 5 Interrelations between emotion categories.

6 Spearman rank correlation coefficients among the study variables are depicted in 7 Table 3. As expected, N- negatively correlated with pleasant experiences, and positively 8 correlated with total unpleasant experiences. N+ negatively correlated with pleasant and 9 dysfunctional experiences and positively correlated with total unpleasant and strong unpleasant experiences. P+ correlated negatively with P- (functionally opposite effect), 10 total unpleasant and dysfunctional enotions. P+ was also positively correlated with total 11 12 pleasant experiences. Finally, P- correlated negatively with functionally helpful and strong pleasant experiences, and positivel with functionally harmful experiences. 13 14 Interestingly, only N- emotions correlated significatily and negatively with shooting scores. Figure 4 depicts the minimum spanning tree based on the maximum 15 correlational path principle. A first major cluster was characterized by pleasant (helpful 16 and harmful) emotions. The second major cluster was formed by a pleasant (helpful 17 18 and harmful) emotions.

1	Discussion
2	This study aimed to explore the interactive effects of multiple emotions on athletic
3	performance in three elite skeet shooters using the aggregated ESP-40 scale. We
4	hypothesised that recalled emotional experiences accompanying successful and poor
5	performances could be instrumental in prediction of actual performances. Typical
6	success profiles were obtained before "better than standard" performance as well as
7	before personally acceptable "standard" performances. These two "personal successful
8	performance" profiles were compared with the emotional profile of sub-standard
9	performances.
10	Recalled best and worst performances
11	Our findings provide partial emetrical support for the notion that multiple emotions
12	(positively- and negatively-toned) have edaptational significance in their co-occurrence
13	before and during task execution. Previous research focusing on assessment of single
14	and discrete emotions was not focused on the interactive effects of different emotions.
15	Our findings, although based on single-case studies, provide empirical support for the
16	assumption that co-occurrence of different emotions is manifested in multiple appraisals
17	inducing pre-event anticipatory emotions. These include challenge-elated (P+) and
18	benefit-related emotions and moderate threat-related (N+ emergency, emotions. In
19	contrast with previous research relatively high level of benefit-related (positive
20	outcomes and gain) emotions were not always detrimental to performance.
21	The absence of dejection-related emotions (N-) was observed in all three levels
22	of performances (Fig.1) and co-occurring with the predominance of challenge-related
23	emotions. On the other hand, the presence of positively-toned emotions (P-) before all
24	three most successful competitions suggests that this pattern is consistent. This also

1	indicates a favourable condition during preparation for these events. High scores of P-
2	in combination with challenge (P+) and moderate level in emergency (threat-related
3	N+) indicates that interactive effects provide more substantial information and higher
4	predictive validity. In support of this assumption the data on recalls of the three
5	unsuccessful competitions supports the success-related profiles: the lowest scores were
6	in P+ category (challenge-related) with moderate scores in emergency $(N+)$ and
7	dejection (N-) emotion category. The athlete was not quite ready for the competition but
8	still maintained positive mindset in all these poor competitions.
9	Recall measures and prediction of actual performance
10	Recall measures were used to estimate the predictive validity of actual performances
11	(Fig.2). Each athlete executed 25 shots which were classified into "better than
12	standard", personally "acceptable standard" and "sub-standard" performance
13	categories. We explored the possibility of asing boxplots as a summary of frequency
14	data and results showed clear coincidence of actual cores with success-related emotion
15	profiles: predominance of P+, slightly lower with P-, the abience of N-, and low on N+.
16	In contrast, a single actual sub-standard performance had a predominant P-
17	(complacency in response to a favourable outcome) in a poor performance emotion
18	profile.
19	Using Ordinal Logistic Regression (OLR)-based estimation of emotion intensity

19 Using Ordinal Logistic Regression (OLR)-based estimation of emotion intensity 20 and probability curves for different performance levels (Kamata et al, 2002) for optimal 21 performance of the same shooter across four emotion categories is less clear in a visual 22 presentation of the same summary of actual scores. One practical problem with using 23 the OLR-based approach to estimate intensity zones is that often there are not enough 24 observations to develop the probability curves (Fig.3). The number of observations

1	required depends on the nature of the data. Although two observations per performance
2	category are the minimum to estimate probability curves, it is important to note that the
3	required observations depend to a large extent on data distribution. Thus, in the case of
4	elite performers were data are narrowly distributed, more observations are needed until
5	one obtains a balanced distribution of emotion intensities associated with all
6	performance categories. Three to five observations per category are usually necessary.
7	In addition, shooters need to perform optimally in current (actual) situations, which may
8	not always be possible. Secondly, the OLR-based procedure is a post-performance
9	(retrospective) method that can be used only after data were collected and its predictive
10	validity still needs to be further examined. This method is actually a summary of
11	available frequency data to cate sorive multiple scores but does not provide the criteria
12	for predictions of forthcoming performances. Moreover, probability curves for all four
13	categories can only be developed separately for each emotion and they do not capture
14	the co-occurrence of four emotion categories. Boyp of representations of multiple
15	emotions data seems an adequate option. Further research requires description and
16	testing the validity of probability curves in prediction of for a coming performance. It is
17	important to clarify if probability curves need to be identified on each occasion or
18	whether they can be used across several competitions once identified,
19	Correlational data (Table 3 & Fig.4 across three shooters N=60 observations)

suggest that there are several types of interactive effects across different constellations between valence and functionally – the same or contrast impact positively-toned with positively toned (by increasing the total valence effects, or functionally predominant). Interestingly, only N- emotions (weak and de-motivational category reflecting a lack of resources) correlated negatively with performance scores. The other emotions had apparently only an indirect impact on performance. Although data presented here was
 correlational, thus, not implying a causal link, this line of research may be good to
 pursue in the future.

4 Our findings suggest that recalled optimal and dysfunctional performances and 5 related emotional experiences as assessed by the ESP-40 scale could be used in 6 prediction of multiple current assessments. These results are also in line with the 7 previously formulated assumptions that the prediction of performance should be based 8 on the assessment of interactive effects rather than on separate emotions (Hanin, 2004, 2007). Interactive effects include the contrasts between and within four emotion 9 10 categories and across eight form modalities of the psychobiosocial (PBS) state (Ruiz, Hanin, & Robazza, 2011). Is op study, positively-toned helpful emotions (P+) and 11 negatively-toned harmful emotions (A-) respect to be the core categories that co-occur 12 and through this interaction determine successful or unsuccessful performance. 13 14 It is important to note that ESP-40 (10 iteras in each of the four categories) is a basic form of the scale. However, to make the scale more personal and relevant for the 15 athlete a shorter version for repeated assessment can be developed using the best five 16 core items (ESP-20), or the best three core items (ESP-12) in each category. 17 This exploratory study includes three case studies, which implies that generalization of 18 19 the results must be taken with caution. However, according to generalizability theory, in 20 the case of idiographic approaches applied to elite level athletes, the estimation of 21 emotional patterns based on several observations from one or few individuals also 22 allows for generalization of findings (Hagtvet & Hanin, 2007). 23 One limitation of this study was that in recall of three "best-ever" and three

24 "worst-ever" competitions, performance as a task execution process was not assessed.

- 1 Moreover, in the assessment of current performance, only outcomes were measured.
- 2 Although in this study it was not possible to assess individual patterns in task execution,
- 3 the action-centered profiling would be most relevant and a promising research direction
- 4 in future research of performance-related PBS states (Bortoli et al. 2012; Hanin, 2010a,
- 5 2011; Hanin, & Hanina, 2009; Ruiz, et al 2011).

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RESUMEN: Este estudio investiga los efectos interactivos de múltiples enociones en el rendimiento (resultado) de tres tiradores de skeet de elite. La escala de emociones a regadas Emotional State Profile-40 (ESP-40; Hanin, 2010b) evalúa las experiencias emocionales recordadas asoci das con rendimientos exitosos y pobres y con rendimientos actuales. El rendimiento actual consiste en 20 series de 25 tiros. Los resultados confirman la utilidad práctica de la escala ESP-40 en la evaluación de los recors interactivos de cuatro categorías de emoción en el rendimiento deportivo. Las emociones recordadas fueron instrumentales en la predicción de los resultados actuales. Se discute la utilidad de la regresión logística

- 32 ordinal en la predicción del rendimiento.
- 33

<sup>10</sup> INTERACTIVE EFFECTS OF EVONONS ON PERFORMANCE: AN EXPLORATORY STUDY IN 11 12 ELITE SKEET SHOOTERS *KEYWORDS*: IZOF model, performance prediction, emotion interaction, aggregated scale. ABSTRACT: This study explored in eractive effects of multiple emotions on performance (outcomes) in three male elite skeet shooters. Recalled emotional experiences associated to successful and poor 13 14 15 performances and to current (actual) performances were measured using the aggregated Emotional State Profile-40 scale (ESP-40; Hanin, 2010b). Current performance involved 20 series of 25 shots. The results provide support for the practical utility of ESP 40 in the assessment of interactive effects of four emotion categories on performance. Recalled measures were instrumental in the prediction of current 16 17 18 19 performances. The utility of ordinal logistic regression-based stimation is discussed. 20 21 EFECTOS INTERACTIVOS DE MÚLTIPLES EMOCIONES OBRE EL RENDIMIENTO: ESTUDIO 22 23 EXPLORATORIO DE TRES TIRADORES DE SKEET DE ELITE 24 PALABRAS CLAVE: modelo IZOF, predicción del rendimiento, nteraz lón emocional, escala agregada. RESUMEN: Este estudio investiga los efectos interactivos de múltiples e nociones en el rendimiento

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18	

		Emotion Cate	gories	
Performances	Unpleasant	Unpleasant	Pleasant	Pleasant
	harmful (N-)	helpful (N+)	helpful (P+)	harmful (P-)
Better than standard (n=1)	) 10 (10-11)	20 (19-20)	38 (37-40)	32 (30-32)
Standard (n=8)	10 (10-11)	20 (19-20)	38 (36-40)	32 (30-34)
Below standard (1-1)	12 (N/A)	19 (N/A)	33 (N/A)	36 (N/A)
	や、		(maximum).	

1 Table 1. Median and min-max range (in brackets) for pre-event emotions (shooter A).

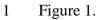
		Emotio	n categories	
Estimation Method	N-	N+	P+	Р-
Recalled method				
Success-related intensity zones	(10, 10)	(20, 20)	(34, 36)	(34, 36)
Failure-related intensity zones	(22, 28)	(24, 27)	(19, 23)	(26, 31)
OLR-based method				
Optimal intensity zones	(10, 11)	(19, 20)	(37, 40)	(30, 33)
Non-optimal intensity zones	(>11)	(<19,>20)	(<37,>40)	(<30,>33
T T				

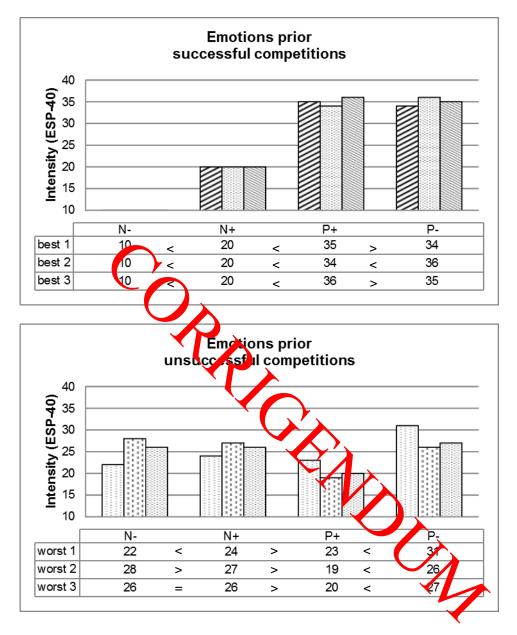
*Table 2.* Lower and upper limits for prediction of optimal performance.

Emotions	1	2	3	4	5	6	7	8	9	10
1. N-										
2. N+	.12									
3. P+	35**	40**								
4. P-	29*	45**	43**							
5. Pleasant	63**	80**	.58**	.43**						
6. Unpleasant	.62**	.81**	58**	42**	-1 **					
7. Helpful	11	.44**	.58**	88**	18	.18				
8. Harmful	.11	42**	59**	.89**	.18	17	99**			
9. Strong unpleasant	34**	.80**	.26*	19	43**	.44**	.41**	39**		
10. Strong pleasant	.07	.11	.16**	88**	02	.01	.89**	90**	01	
11. Shooting score	26*	13	08	.25	.17	16	14	.14	.03	21
Note. Pleasant= P+	+ P-; un	pleasan	t = N +	+ N-;/	he pfu	$l = P + \cdot$	+ N+; h	armful	= P	⊦ N-;
strong unpleasar	nt = N + r	- N-; ar	d stron	g pleas	sant -	P+ - P-	;*p<.	05, ** <sub>]</sub>	p < .01	•
						$\gamma$				
						C				
								1		

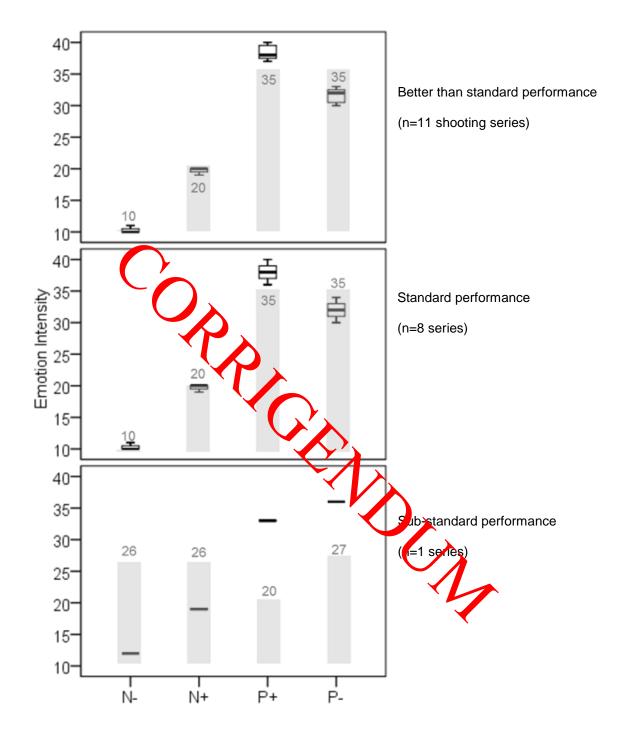
1 Table 3. Intercorrelations between emot	tions before 20 series for 3 shooters.
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1	Figure captions
2	Figure 1
3	Performance emotions before most successful and unsuccessful performances (shooter
4	A).
5	Figure 2
6	Boxplots of emotion intensities prior-current and bar graphs prior-recalled performances
7	(shooter A).
8	Note. Bar graphs and B represent median of emotion intensities before three best
9	performances; bar groons C represents emotion intensities before three worst
10	performances
11	Figure 3
12	Probability curves based on intensities of four emotion categories (shooter A).
13	Note. OP = optimal performance; nOP/Bnor-optimal performance with emotion
14	intensity below average; and nOP/A = non-optimal performance with emotion intensity
15	above average.
16	Figure 4
17	Minimum spanning tree for emotions before performance (n=60 ob ervations).
18	

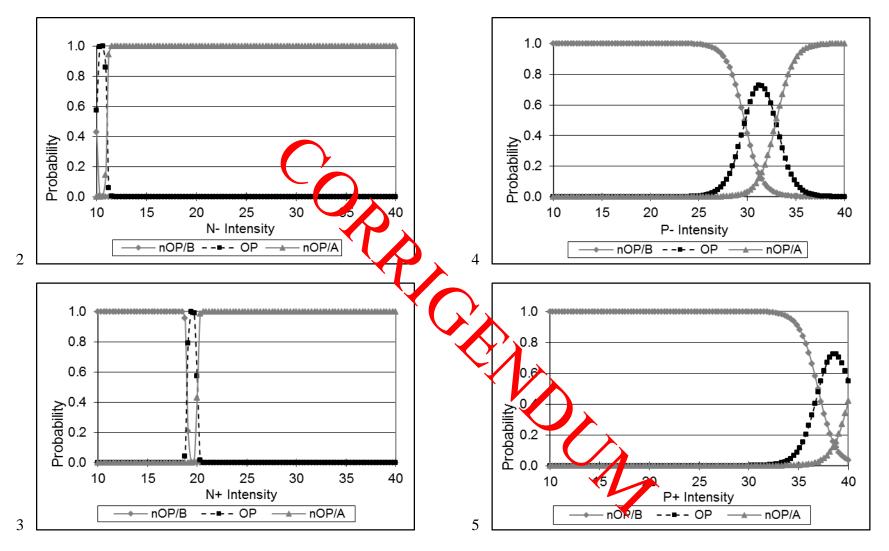




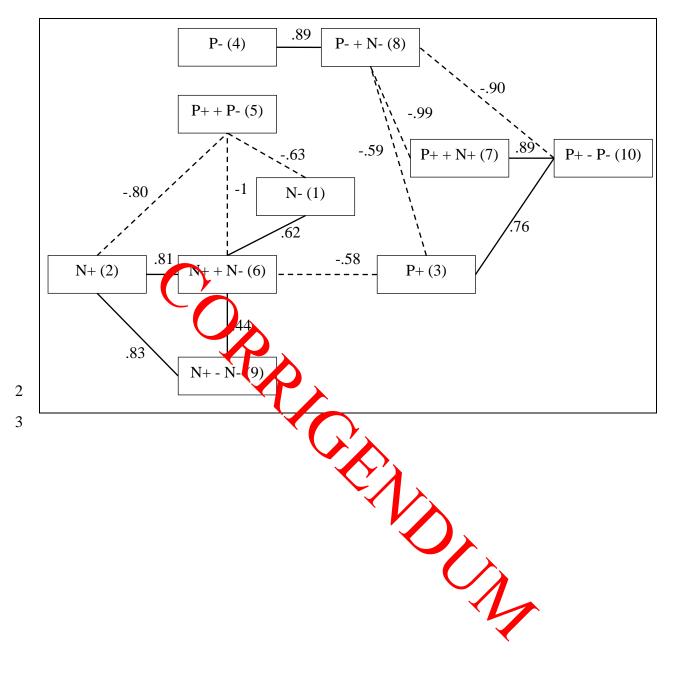




1 Figure 3.

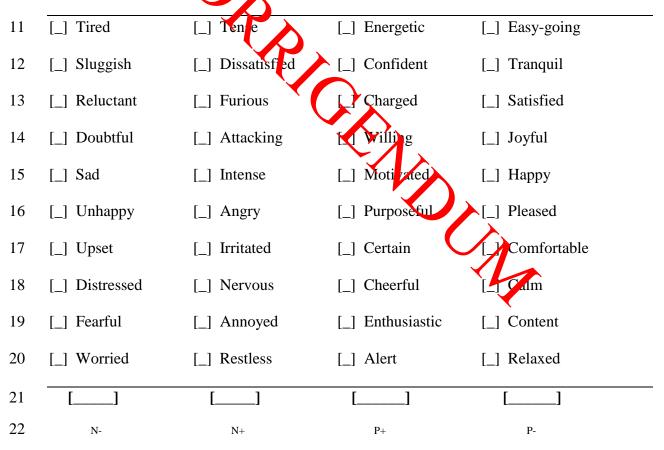


# 1 Figure 4



## 1 EMOTIONAL STATE PROFILE (ESP - 40)

- 2 The ESP helps to describe *how* you think *you feel* in different performance situations. There
- 3 are no right or wrong responses! Make sure you:
- 4 Consider how you actually feel (or felt), not how you would like to feel.
- 5  $\circ$  Work across the page.
- 6  $\circ$  Number the words in each row
- 7 Give a 4 to the word that *best* describes you or that you relate to *best*.
- 8 Give a 3 to the next best, then 2, and then 1 to the *least*.
- 9 Make sure each row has a 4, 3, 2, and 1 (no duplicates)
- 10  $\circ$  Go with your first reaction.



## 23 Instructions for scoring:

- o Add up each column of the scale and put the totals below.
- 25 Plot your scores & connect the points to create a graph