

Mega-analysis of the Interoceptive Accuracy Scale (IAS) Structure and its Subjective Correlates

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Introduction

Interoception refers to the process of sensing, interpreting and integrating information pertaining to internal organs, such as the heart, the lungs or the gut (Khalsa, Adolphs, Cameron, Critchley, Davenport, Feinstein, Feusner, Garfinkel, Lane, Mehling, & others, 2018). While recent research emphasizes a key role of interoception in a variety of

processes (e.g., emotion regulation, decision making) and of outcomes (physical and psychological well being), the field remains clouded by concerns about how interoception is assessed.

The Interoceptive Assessment Puzzle

Various measures of interoception have been developed (see Figure 1), forming a combination of “objective” and “subjective” assessments (i.e., physiological tasks such as the heart beat counting or tracking vs. questionnaires and subjective scales involving metacognitive judgments), “explicit” and “implicit” paradigms (i.e., directing participants’ awareness and attention to interoceptive processes vs. measuring interoception unbeknownst to them), various interoceptive modalities (e.g., cardioception, respiroception, gastroception) and theoretical dimensions (e.g., accuracy, sensitivity, awareness). While there is no consensus as to which particular approach provides the most accurate and “pure” measure of interoception and interoceptive abilities (assuming it is a unidimensional construct), it is instead plausible that each measure has strengths and limitations, and a utility dependent on the context and goal at hand (Desmedt et al., 2023; Jahedi & Méndez, 2014).

Although the use of subjective self-report questionnaires to measure deeply embodied functions might seem paradoxical at first, recent redefinitions of interoception emphasize the role of high-level and metacognitive elaboration of interoceptive information. These redefinitions provide theoretical grounding to support the idea that some facets of interoception, including participants’ metacognitive beliefs, can be assessed subjectively (Khalsa, Adolphs, Cameron, Critchley, Davenport, Feinstein, Feusner, Garfinkel, Lane, Mehling, & others, 2018; Suksasilp & Garfinkel, 2022). Moreover, the notion that self-reports might not reflect the same processes as other interoception tasks might be important to contextualize the apparent lack of convergence between measures in the field (Desmedt et al., 2022). For instance, existing findings

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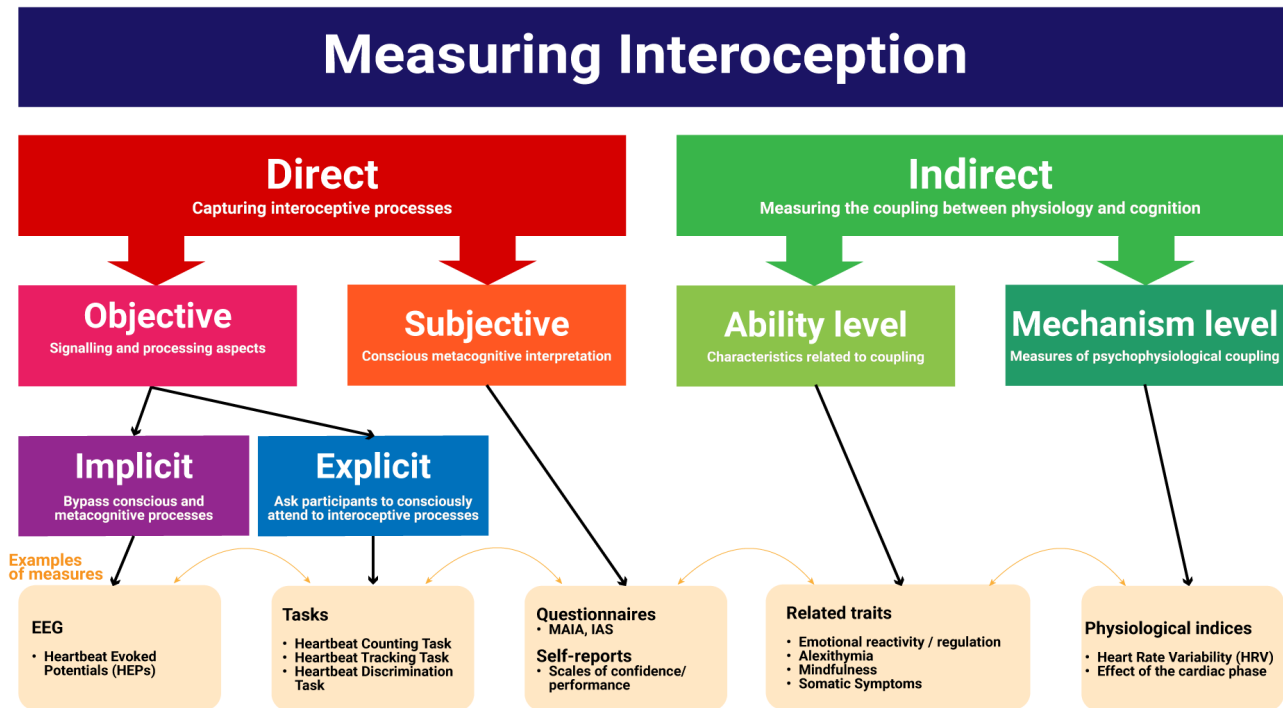


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Figure 1

The Interoceptive Assessment Puzzle. The different modalities of interoception (e.g., cardioception) can be assessed directly or indirectly. Direct assessments can further be subjective or objective, depending on whether they involve conscious metacognitive appraisals or more performance-based indices. Interoceptive tasks can be explicit (the participant is aware of the interoceptive nature of the task and must consciously attend to interoceptive signals; e.g., the heartbeat counting task) or implicit (measurements of interoception done unbeknownst to the participant; e.g., heartbeat evoked potentials measured during resting state). Indirect assessments evaluate constructs typically related (and ideally dependent on) to interoceptive processes or ability (or its deficit).



typically show weak or no correlations between questionnaires and objective measures, such as the Heartbeat Counting Task (HCT, Schandry, 1981) and the Heartbeat Detection Task (HDT, Kleckner et al., 2015), including for measures of the same theoretical dimensions (Arslanova et al., 2022; Brand et al., 2023; e.g., task-based accuracy vs. self-reported accuracy, Murphy et al., 2019). Additionally, even various objective measures assessing in theory the same interoceptive dimension, such as accuracy, either show no or weak correlation (respectively, Brand et al., 2023; Hickman et al., 2020). Perhaps more surprisingly, low correlations have been observed even among questionnaires, suggesting (in parallel to major validity concerns) the potential targeting of different facets related to interoception.

One striking example concerns the assessment of interoceptive sensibility, which is broadly defined as the self-reported tendency to focus on and detect internal sensations (Garfinkel et al., 2015), but more narrowly as the subjective tendency to focus on interoceptive signals, without necessar-

ily implying detection ability (Khalsa, Adolphs, Cameron, Critchley, Davenport, Feinstein, Feusner, Garfinkel, Lane, Mehling, & others, 2018). A recent systematic review suggested that various questionnaires designed to assess interoceptive sensibility may, in fact, measure distinct constructs, with the risk of researchers treating them as equivalent despite overall low convergence (Desmedt et al., 2022). Notably, this review adopted a broad definition of sensibility, incorporating both interoceptive sensibility and interoceptive self-report scales, following the eight-facet model by Khalsa, Adolphs, Cameron, Critchley, Davenport, Feinstein, Feusner, Garfinkel, Lane, Mehling, and others (2018). Several widely used questionnaires were included in the review, such as the Multidimensional Assessment of Interoceptive Awareness (MAIA, Mehling et al., 2012; MAIA-2, Mehling et al., 2018), the Body Perception Questionnaire (BPQ, Porges, 1993), the Private subscale of the Body Consciousness Questionnaire (PBCS, Miller et al., 1981), the Body Awareness Questionnaire (BAQ, Shields et al., 1989), and the Eating

Disorder Inventory (Garner et al., 1983; EDI, Garner, 1991). The lack of correlations to moderate correlations among these questionnaires highlight the need for greater conceptual clarity regarding what each measure captures, how they relate to different dimensions of interoception, and their potential overlaps with other constructs, such as alexithymia and body awareness.

The Interoceptive Accuracy Scale (IAS)

Focusing on the *accuracy* dimension of interoception, a recently developed scale with a rapidly growing popularity is the Interoceptive Accuracy Scale (IAS, Murphy et al., 2019). The IAS consists of 21 Likert-scale items that query how accurately one can perceive different bodily signals, with one item per physiological modality such as respiration (“*I can always accurately perceive when I am breathing fast*”), heart (“*I can always accurately perceive when my heart is beating fast*”), skin (“*I can always accurately perceive when something is going to be ticklish*”), arousal or bodily functions like coughing (“*I can always accurately perceive when I am going to cough*”) or urinating (“*I can always accurately perceive when I need to urinate*”). Appealingly, the IAS’ statements are about specific interoceptive behaviours and “objective” experiences (although still subjective reports of them), which is a distinct difference with other popular interoception questionnaires, such as the MAIA-2, which contains more general and metacognitive items (e.g., “*I trust my body sensations*”, “*I can notice an unpleasant body sensation without worrying about it*”), as well as dimensions related to attention regulation (e.g., Not-distracting) or emotion regulation (e.g., Not-worrying).

The original validation study suggested a two-factor structure for the IAS, one reflecting the perception of general interoceptive signals (urinate, hungry, defecate, thirsty, pain, heart, taste, breathing, temperature, muscles, affective touch, vomit, sexual arousal), and other relating to signals that may be difficult to perceive solely through interoceptive information (itch, tickle, cough, burp, bruise, blood sugar, sneeze, wind). The authors however underlined its acceptable but imperfect fit (Murphy et al., 2019, p. 127), and several follow-up studies have indeed identified different optimal solutions. For instance, Brand et al. (2023) reported a 1-factor solution, while Lin et al. (2023) and Campos et al. (2021) found bifactor solutions (i.e., one general factor above a set of lower-level factors, Rodriguez et al., 2016) to be the best fit. Using a 2-factors Exploratory Factor Analysis (EFA), Koike and Nomura (2023) suggested that the items could be grouped into cutaneous (itching, tickling, coughing, burping, affective touch, bruising, passing gas, sneezing, muscle sensations, sexual arousal, and taste) and visceral sensations (urination, defecation, hunger, thirst, pain, breathing, fatigue/blood sugar, temperature, vomiting, and heartbeat).

Discussions have also been focused on specific items.

For instance, Murphy et al. (2019) notes that some items might measure direct interoceptive signals such as cardioception, while others might capture phenomena not perceivable through interoceptive signals alone (e.g., “bruising”; p. 119). Lin et al. (2023) additionally highlights five locally dependent pairs and three items (touch, blood sugar, bruise) with exceptionally high difficulty and low discrimination, and Campos et al. (2021) reported “tickle” to be the only item that reflected more specific factors than the general factor. Furthermore, localization issues also arose, with both “itch” and “tickle” corresponding to the same Chinese character, leading to their collapse into a single item (Lin et al., 2023).

Regarding its validity, the IAS has naturally been compared to other interoception-related measures, and shows a positive correlations with most facets of the MAIA (Gagero et al., 2021; Mehling et al., 2018), except for the Not-Distracting and Not-Worrying subscales (Brand et al., 2023) - which were previously highlighted as related to non-interoceptive abilities (Ferentzi et al., 2021). Interestingly, findings on the correlation between the IAS and the body awareness dimension of the BPQ (i.e., BPQ-A) have been mixed: some studies report small positive correlations (Brand et al., 2023; Campos et al., 2021; Koike & Nomura, 2023), while others find small negative correlations (Lin et al., 2023) or no correlation at all (Murphy et al., 2019). Positive correlations have also been observed with the “observation” and “description” subscales of the Five Facet Mindfulness Questionnaire (FFMQ, Baer et al., 2006; Brand et al., 2023; Koike & Nomura, 2023), as well as with the “non-reactivity” and “acting with awareness” subscales (Koike & Nomura, 2023). Additionally, the IAS has shown a positive correlation with the interoceptive awareness subscale of the Eating Disorder Inventory (EDI-IA, Lin et al., 2023) and a negative correlation with the Interoceptive Confusion Questionnaire (ICQ, Brewer et al., 2016), as reported by Brand et al. (2023) and Murphy et al. (2019). Lastly, the correlation with the Interoceptive Attention Scale (IATS, Gabriele et al., 2022) appears rather small (Koike & Nomura, 2023; Lin et al., 2023).

While assessing the predictive validity of an interoception scale can be conceived as theoretically challenging, expected negative associations were observed between the IAS and alexithymia (Brand et al., 2023; Campos et al., 2021; Koike & Nomura, 2023; Lin et al., 2023; Murphy et al., 2019), somatic symptoms (Brand et al., 2023; Koike & Nomura, 2023; Lin et al., 2023), depressive symptoms (Brand et al., 2023; Koike & Nomura, 2023; Lin et al., 2023), anxiety (Brand et al., 2023), neuroticism (Brand et al., 2023) and self-esteem (Murphy et al., 2019). Taken together, these findings support the IAS as measuring an adaptive aspect of interoception, although its pattern of associations with other interoception (or interoception-related) questionnaires points towards some overlap across various theoretical dimensions, casting some doubt on the orthogonal models of interoception and the pos-

sibility of its faithful capture by questionnaires.

Given the increasing relevance of interoception across psychology and neuroscience, and the growing use of the IAS as an interesting self-report measure of interoceptive accuracy, the current study aims at 1) clarifying the structure of the IAS with a mega-analytic (which involves a re-analysis at the raw data level by aggregating datasets) approach that leverages existing data and contrast the traditional CFA/SEM factor-based analyses with network-based ones (Exploratory Graph Analysis); 1) provide an overview of the dispositional correlates of the IAS, clarifying its general pattern of associations, which is key to better understand the nature, place and role of interoception questionnaires within a larger context.

Study 1

Study 1 will re-analyse and assess the factor structure of the IAS by taking advantage of the large number of open-access datasets (Arslanova et al., 2022; Brand et al., 2022; Brand et al., 2023; Campos et al., 2021; Gaggero et al., 2021; Lin et al., 2023; Murphy et al., 2019; Poerio et al., 2024; Todd et al., 2022; Von Mohr et al., 2023). While combining these studies might provide a more robust and generalizable understanding of the IAS' factor structure, we also additionally provide an individual analysis (i.e., applying the same method on all samples separately) to add nuance to the general picture, as all studies differ in their sample size, demographic characteristics, language, and procedure.

Methods

Datasets

Our search focused on studies citing the original IAS validation paper (Murphy et al., 2019), identifying 136 papers (as of 01/05/2024). To qualify for inclusion, papers needed to (1) provide accessible data in open-access, (2) employ the IAS as a measure, and (3) report individual IAS items scores. We also included the data of five unpublished studies. A total of 16 datasets was included (see **Table 1**).

The total number of participants was 33,272 participants (Mean = 48.09 ± 13.1, 71.3% Female).

Data Analysis

Psychometrically good items should exhibit various qualities, such as validity, reliability and discrimination, to which one of the contributing factors is the amount of inter-individual variability captured by an item. Items to which all participants' answers are concentrated around one option - i.e., exhibiting a narrow distribution - will be flagged as potentially problematic.

After examining the distributions and correlations of all IAS items, we will test for "redundant" items (e.g., due to multicollinearity or local dependency) using Unique Variable Analysis (UVA, Christensen et al., 2023), a novel method

derived from network psychometric designed to identify and merge items that share substantial variance (which can distort the structure estimation). We will use a threshold of 0.30 that detects large to very large overlap to remain conservative and only suggest scale modifications if strongly justified.

Following the analysis of items, we will analyze the factor structure of the IAS using three different approaches, each with particular trade-offs and assumptions, to provide a multi-verse picture of likely solutions. Namely, we will apply traditional exploratory and confirmatory Factor Analysis (EFA/CFA), hierarchical clustering (HCA), and Exploratory Graph Analysis (EGA), to the whole sample, as well as to each dataset separately (details being available in the analysis document at <https://github.com/RealityBending/InteroceptionIAS>), and our decisions and conclusions will try to take into account both levels of the analysis.

By combining network analysis with psychometric methods, the recently-developed Exploratory Graph Analysis (EGA, H. F. Golino & Epskamp, 2017) framework allows to jointly estimate the number of dimensions (i.e., groups of items), the structure as well as its stability (H. Golino et al., 2020; H. F. Golino & Epskamp, 2017). Evidence has underlined its suitability as an alternative to traditional factor analysis, addressing some of its limitations such as the assumption of a "latent" source of variability, possible biasing in the estimation of the optimal factor numbers depending on sample size, and the poor performance of other methods in complex population structures, while remaining comparable and interpretable (Christensen & Golino, 2021b; Jiménez et al., 2023). At a fundamental level, EGA conceptualizes variables as nodes in a network, with connections (edges) reflecting associations between them. Clustering these nodes reveals distinct communities of related items, in practice akin to traditional latent factors - but without explicitly assuming their presence (Christensen & Golino, 2021b). We used the EGAnet package (Christensen & Golino, 2021a) to fit a hierarchical EGA with the Leiden community detection algorithm.

While EGA offers a robust alternative to traditional factor analysis, factor analysis remains a widely used method for dimensionality assessment. As our goal is to provide a general - yet nuanced - picture, with room to show potential discrepancies emerging from the methods used, we will also include it in the present study. Unlike EGA, factor analysis assumes that a latent source of variability — a common latent variable — underlies the observed set of manifest variables (Cosemans et al., 2022). A critical step in factor analysis is determining the optimal number of factors, for which we will use the Method Agreement Procedure (Lüdtke et al., 2021), which involves a consensus-based decision based on multiple factor estimation methods applied concurrently.

Finally, we will also apply Hierarchical Clustering Analysis (HCA, Murtagh & Legendre, 2014), which differs from

Table 1*Description of the samples used in the study.*

Sample	Reference	Language	N	Difference	Age (Mean \pm SD)	Range	Female %	Availability
Murphy et al., (2020)								osf.io/3m5nh
Sample 1a		English	451		25.8 \pm 8.4	18-69	69.4%	
Sample 1b		English	375		35.3 \pm 16.9	18-91	70.1%	
Sample 2	Gaggero et al., (2021)	English and Italian	814		24.9 \pm 5.3	18-58	60.3%	osf.io/5x9sg
Sample 3	Campos et al., (2022)	Portuguese	515		30.7 \pm 10.5	18-72	59.6%	osf.io/j6ef3
Sample 4	Todd et al., (2022)	English	802		48.6 \pm 14.1*	18-92*	50%*	osf.io/ms354
Sample 5	Arslanova et al., (2022)	English	143		28.5 \pm 7.6	18-73	46.8%	osf.io/mp3cy
Sample 6	Brand et al., (2022)	German	619		43.9 \pm 14.5	18-78	78.7%	osf.io/xwz6g
	Brand et al., (2023)							osf.io/3f2h6
Sample 7a		German	522		23.4 \pm 6.7	18-79	79.5%	
Sample 7b		German	1993		32.0 \pm 12.6	16-81	77.7%	
Sample 7c		German	802		27.3 \pm 9.3	18-72	68.9%	
Lin et al., (2023)								osf.io/3eztd
Sample 8a		Chinese	1166	Collapsed "Itch" and "Tickling"	32.5 \pm 8.4	16-60	57.0%	
Sample 8b		Chinese	500	Collapsed "Itch" and "Tickling"	37.4 \pm 7.4	20-60	56.2%	
Sample 9	VonMohr et al., (2023)	English	21843		56.5 \pm 14.4	18-93	73.2%	osf.io/7p9u5
Sample 10	Makowski et al., (2023a)	English	485	Analog scales. No Temperature, Blood sugar and Cough items	30.1 \pm 10.1	18-73	50.3%	github.com/RealityBending/IlusionGameReliability
Sample 11	Makowski et al., (2023b)	English	836	Analog scales	25.1 \pm 11.3	17-76	53.0%	github.com/DominiqueMakowski/PHQ4R
Sample 12	Makowski et al., (2023c)	English	146	Analog scales	21.1 \pm 4.3	18-50	76%	github.com/RealityBending/InteroceptionPrimals
Sample 13	Makowski et al., (2024)	English	737		36.8 \pm 14.9	17, 87	57.3%	github.com/RealityBending/InteroceptionScale
Sample 14	Poerio et al., (2024)	English	107		26.8 \pm 9.2	18-57	74.8%	osf.io/49wbv
Sample 15	Poerio et al., unpublished	English	131		30.9 \pm 12.0	18-60	75.9%	
Sample 16	Arjona et al., unpublished	English	279		26.4 \pm 13.2	18-79	67.7%	
Sample 17	Total		33272		48.09 \pm 13.1	17-93	71.3%	

*Information taken from the sample description of relevant paper rather than recomputed.

factor analysis in that it does not assume any latent source of variability, but instead iteratively groups items based on their similarity (e.g., correlation) into a hierarchy of clusters. The benefits of HCA include its interpretability and ability to capture complex relationships among items without relying on strict assumptions about data distribution or latent variables.

In a typical 2-step fashion, this first analysis run will be followed by a structure refinement with a further selection of items, and the final pool of items will be tested again. Additionally, various solutions (e.g., adding general factors) will be statistically compared using Confirmatory Factor Analysis (CFA).

Results

The distribution of the items across samples suggests the presence of a consistent modal value (Figure 2). In other words, participants are most likely to answer 4/5 (i.e., agree) on most items, with the exception of "blood sugar" and "bruise", which exhibit a different distributional pattern with a lower mode (~2/5). While it is not a problem *per se*, the con-

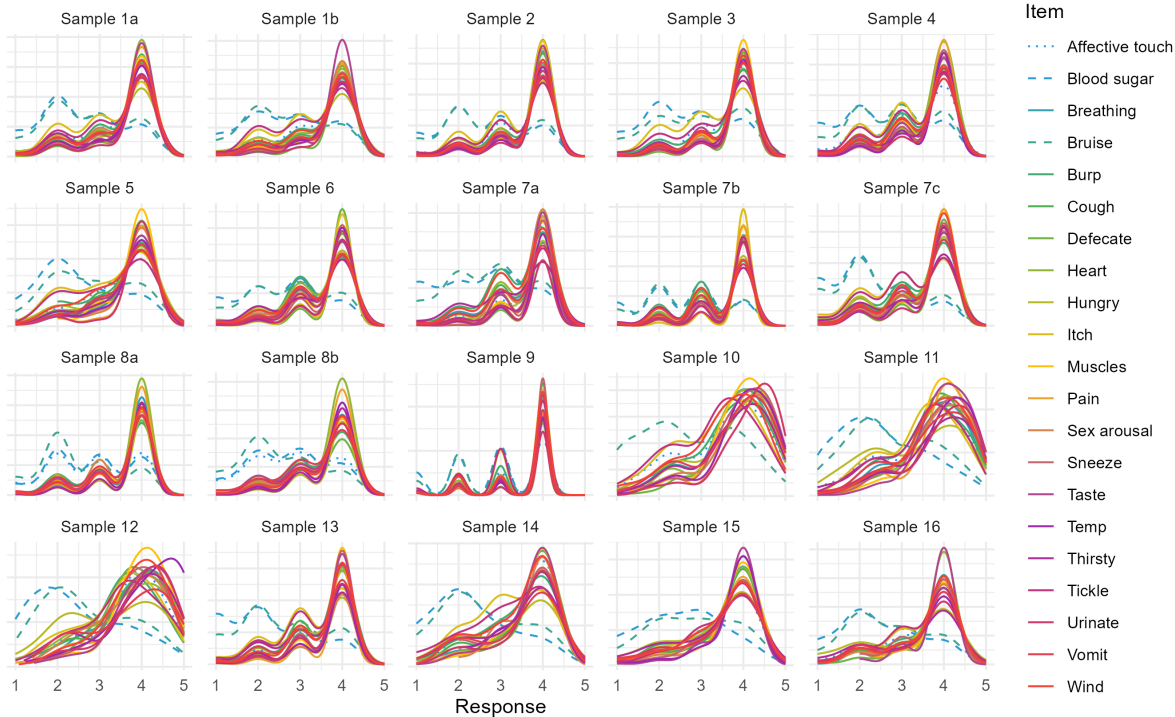
trasting distribution might be indicative of items with non-homogeneous psychometric "difficulty". This is also the case for "affective touch" in samples 8a and 8b (the Chinese validation samples), which might indicate localisation issues. Additionally, one can note the low occurrence of extreme values (1 and 5), meaning that the bulk of answers varies between 3 values (assuming the IAS is implemented as a 5-point Likert scale following its validation). The samples using an analogue scale (samples 10, 11 and 12 in the figure) display a more continuous and progressive spread of answers, seemingly improving the interindividual variability, although potentially displaying a secondary lower mode at ~2 (which might be suggesting the existence of potential clusters of participants). The correlation matrix between all items shows an overall positive correlation pattern, with highly correlated pairs of items (e.g., Tickle-Itch, Urinate-Defecate, Pain-Wind, Hungry-Thirsty) or triplets (Vomit-Sneeze-Cough, Temperature-Muscles-Pain).

UVA flagged two strongly redundant variables, "itch" and "tickle" - suggesting to remove the latter. Several more pairs of items were flagged as moderately redundant ("wind"

Figure 2

Top: Distribution of responses across datasets reveals a consistent modal value, typically around 4 or 5 (indicating agreement), except for ‘blood sugar’ and ‘bruise’ - in dashed lines - and ‘affective touch’ (dotted lines) in the Chinese validation sample, which have lower modes. Most responses cluster around the middle values, with few extreme scores (1 and 5). Samples using an analogue scale (10a, 10b, 10c) show a more continuous distribution and increased interindividual variability. Since most samples use Likert scales (discrete), density plots may not be the most accurate representation but were chosen to clearly highlight variability patterns in the data. Bottom: The correlation matrix between all items shows an overall positive correlation pattern, with correlated pairs (e.g., Wind, Burp) of items or triplets (e.g., Vomit, Sneeze and Cough).

Item Distribution



Correlation Matrix

N = 33272

Breathing	.17	.18	.25	.21	.33	.27	.32	.33	.33	.37	.34	.26	.28	.30	.31	.30	.32	.36	.36	.47	
Heart	.19	.20	.24	.24	.25	.22	.23	.24	.25	.29	.24	.23	.25	.25	.25	.26	.26	.26	.22		.47
Thirsty	.16	.16	.21	.19	.29	.22	.40	.34	.33	.30	.33	.25	.25	.26	.25	.25	.27	.49		.22	.36
Hungry	.17	.15	.21	.21	.30	.22	.35	.32	.33	.28	.33	.24	.23	.25	.24	.24	.24		.49	.26	.36
Cough	.27	.30	.32	.24	.33	.27	.31	.34	.41	.39	.34	.44	.49	.34	.43	.59		.24	.27	.26	.32
Sneeze	.27	.28	.29	.21	.34	.28	.33	.36	.38	.36	.33	.40	.43	.34	.48		.59	.24	.25	.26	.30
Vomit	.22	.23	.29	.22	.36	.30	.33	.39	.35	.34	.33	.34	.37	.35		.48	.43	.24	.25	.25	.31
Taste	.24	.24	.26	.21	.32	.30	.31	.33	.35	.36	.33	.31	.31		.35	.34	.34	.25	.26	.25	.30
Burp	.26	.27	.35	.24	.34	.26	.29	.32	.34	.40	.31	.62		.31	.37	.43	.49	.23	.25	.25	.28
Wind	.25	.25	.31	.23	.35	.25	.31	.34	.34	.37	.30		.62	.31	.34	.40	.44	.24	.25	.23	.26
Pain	.24	.24	.38	.24	.35	.33	.35	.34	.44	.48		.30	.31	.33	.33	.33	.34	.33	.33	.24	.34
Muscles	.26	.27	.37	.26	.37	.32	.33	.35	.42		.48	.37	.40	.36	.34	.36	.39	.28	.30	.29	.37
Temp	.23	.24	.28	.21	.40	.32	.38	.38		.42	.44	.34	.34	.35	.35	.38	.41	.33	.33	.25	.33
Defecate	.20	.19	.22	.17	.36	.28	.59		.38	.35	.34	.34	.32	.33	.39	.36	.34	.32	.34	.24	.33
Urinate	.18	.16	.22	.16	.34	.27		.59	.38	.33	.35	.31	.29	.31	.33	.33	.31	.35	.40	.23	.32
Affective touch	.37	.31	.28	.25	.36		.27	.28	.32	.32	.33	.25	.26	.30	.30	.28	.27	.22	.22	.22	.27
Sex arousal	.21	.18	.22	.18		.36	.34	.36	.40	.37	.35	.35	.34	.32	.36	.34	.33	.30	.29	.25	.33
Blood Sugar	.24	.28	.38		.18	.25	.16	.17	.21	.26	.24	.23	.24	.21	.22	.21	.24	.21	.19	.24	.21
Bruise	.29	.34		.38	.22	.28	.22	.22	.28	.37	.38	.31	.35	.26	.29	.29	.32	.21	.21	.24	.25
Itch	.61		.34	.28	.18	.31	.16	.19	.24	.27	.24	.25	.27	.24	.23	.28	.30	.15	.16	.20	.18
Tickle		.61	.29	.24	.21	.37	.18	.20	.23	.26	.24	.25	.26	.24	.22	.27	.27	.17	.16	.19	.17

and “burp”; “urinate” and “defecate”) and mildly redundant (“sneeze” and “cough”; “heart” and “breathing”; “hungry” and “thirsty”). These patterns consistently appeared in most samples when considered individually. We removed “tickle” from further analysis due to its high redundancy (and because it is absent from some datasets due to translation issues).

The HCA highlighted pairs and triplets of items consistently grouped together across samples, such as “wind” and “burp”, “sneeze” and “cough”, “itch” and “bruise”, “urinate” and “defecate”, and “pain”, “muscles”, and “temperature”. This pattern was largely replicated by the EGA, with the additional presence of a unique cluster comprising “Sex arousal”, “Affective touch”, “Temperature”, “Pain”, “Muscles”, and “Taste”. EFA suggested the optimal number of factors to be 3, yielding one dimension with expulsion-related items (“burp”, “wind”, “cough”, “sneeze”, and “vomit”), a second dimension with viscerosensitive items (“heart”, “breathing”, “hungry”, “thirsty”, “urinate”, and “defecate”), and a third dimension with skin-related items (“bruise” and “blood sugar”).

Importantly, this initial structure analysis run highlighted some problematic items: “taste” typically displayed a lone or unstable pattern of associations, “affective touch” exhibited cross-loadings and instability, “vomit” was less strongly associated with other items, and “itch” did not form a consistent cluster. Finally, “temperature” and “sexual arousal” showed redundant patterns of associations but were less reliable. These 6 items were thus removed, and a second run of structure analysis was performed on the remaining 14 items.

HCA and EGA yielded highly consistent results, emphasizing pairs of items, namely Hungry-Thirsty, Bruise-Blood sugar, Urinate-Defecate, Muscles-Pain, Breathing-Heart, Cough-Sneeze, Wind-Burp. HCA also significantly grouped the Urinate-Defecate and Muscles-Pain pairs, as well as expulsion items (Wind-Burp and Cough-Sneeze). EFA suggested once again 3 factors as the optimal solution, with the first factor including expulsion-related items (“burp”, “wind”, “cough”, “sneeze”), the second factor being related to the Urinate-Defecate pair, and the third factor comprising the remaining items.

We then fitted and compared using CFA various candidate structures emerging from the previous analyses, including a 1-factor model (the G-model), a 3-factor model (EFA), a 3+1 model (EFA + general factor), a 5-factor model (HCA), a 5+1 model (HCA + general factor), a 7-factor model (EGA), and a 7+1 model (EGA + general factor). The EGA model with 7 factors of item pairs provided the best fit with the lowest RMSEA (0.026), lowest χ^2 (72.825), highest CFI (0.988). It was followed by the EGA + general factor model (RMSEA = 0.059, χ^2 = 180.403, CFI = 0.065), and the HCA model with 5 factors (RMSEA = 0.062, χ^2 = 182.492, CFI = 0.921). The other models performed poorly, with RMSEA > 0.08 and CFI < 0.90. The unique-factor model yielded the lowest BIC (which takes into account the number of factors), followed by

the EGA model. All the other models displayed significantly (BIC-based Bayes Factor < 1/100) lower evidence compared to the EGA model.

This result was relatively consistent across unique datasets with the EGA model providing the best fit. In all samples where CFA converged, it showed excellent performance with low RMSEA (\approx 0.025–0.052), high CFI (\geq 0.95), and substantially lower χ^2 compared to alternative models. Adding a general factor systematically worsened fit, while hierarchical clustering and EFA-based models showed poorer performance (RMSEA > 0.05, CFI < 0.95). Although BIC values occasionally favored alternative specifications, Bayes factor comparisons overwhelmingly supported the EGA solution, confirming its robustness across datasets.

Discussion

Study 1 aimed to systematically evaluate the structure of the IAS. Previous research reported conflicting models, from unidimensional (Brand et al., 2023) to bifactor (Campos et al., 2021; Lin et al., 2023) or two-factor solutions (Koike & Nomura, 2023; Murphy et al., 2019). Concerns about specific items (e.g., low discrimination, local dependencies, and items reflecting phenomena less accessible to interoception) prompted a mega-analytic approach using both factor-analytic and network-based methods.

Analyses revealed that seven item pairs (e.g., Hunger-Thirst, Urinate-Defecate, Cough-Sneeze) provided a robust and replicable structure, leading to a refined 14-item IAS. These pairs consistently emerged across datasets and yielded superior fit indices compared to alternative models. This suggests that the IAS does not measure a broad latent construct of interoceptive accuracy but reflects clusters of tightly coupled items tied to specific bodily sensations.

The presence of redundant or unstable items (e.g., Tickle, Taste, Affective Touch) undermines interpretability and likely explains inconsistencies in prior factor-analytic findings (Campos et al., 2021; Lin et al., 2023). Nevertheless, the structured item pairs correspond to intuitive bodily domains—visceral, expulsion, musculoskeletal—supporting ecological validity. These findings highlight the need for targeted scale refinement and motivate future development of context-sensitive, multidimensional interoception measures.

In summary, study 1 suggests that once problematic items are removed, the IAS is best conceptualized as a set of item clusters rather than a unidimensional scale. This structural insight provides the foundation for Study 2, where dispositional correlates are examined to further evaluate the IAS’s construct validity.

Study 2

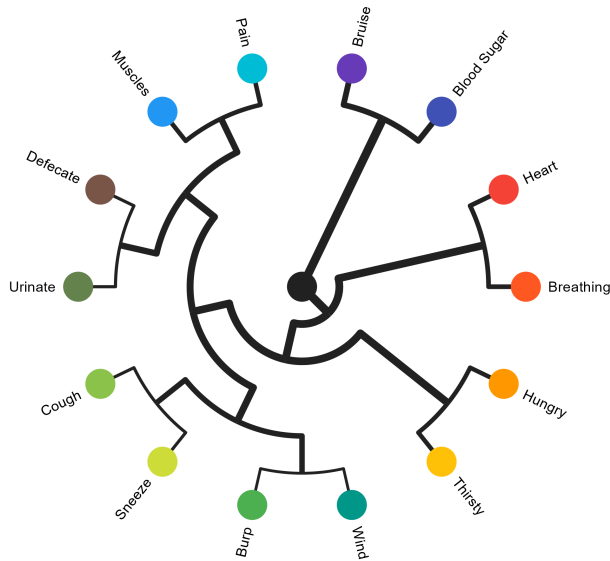
Building on the structural findings from Study 1, the second study turns to the validity of the IAS by examining its dispositional correlates. Although the IAS has been widely used,

Figure 3

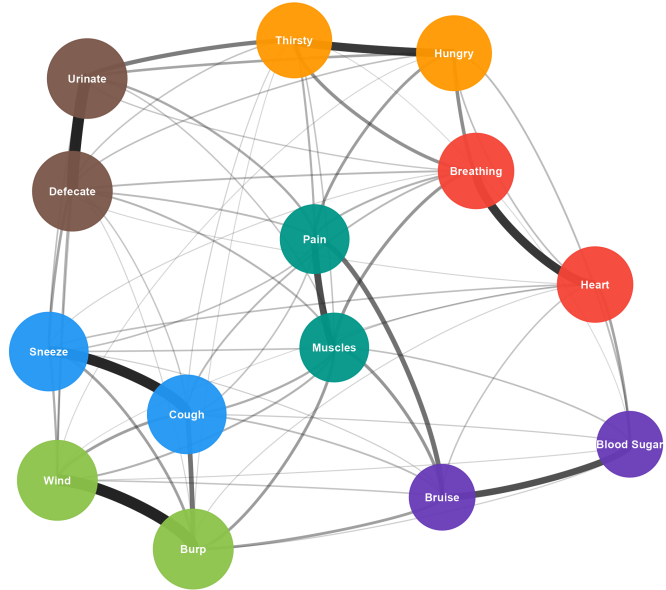
Four structure analysis methods (HCA, EGA, EFA, CFA) were applied and converged on a consistent optimal solution of 14 items formed seven pairs: Hungry-Thirsty, Bruise-Blood sugar, Urinate-Defecate, Muscles-Pain, Breathing-Heart, Cough-Sneeze, Wind-Burp. While EFA suggested 3 factors, CFA confirmed the superiority of the 7-factor model over alternative structures.

Hierarchical Clustering Analysis (HCA)

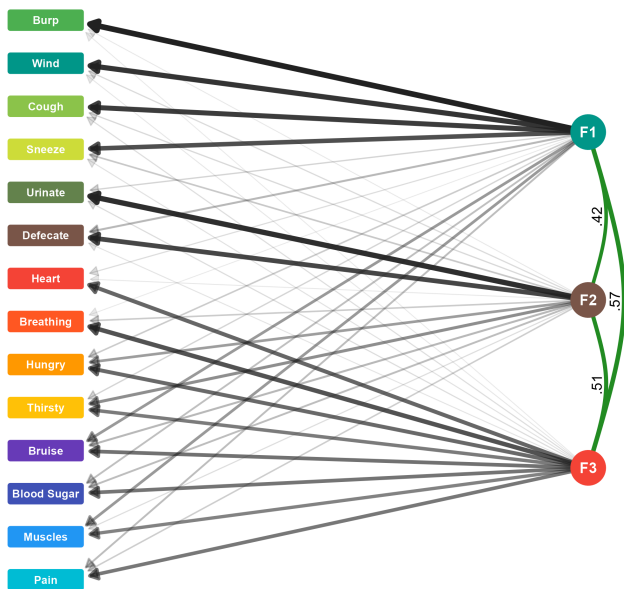
Method = Correlation

**Exploratory Graph Analysis (EGA)**

Method = Leiden

**Exploratory Factor Analysis (EFA)**

Method = Oblimin

**Confirmatory Factor Analysis (CFA)**

Method = Maximum Likelihood



evidence regarding its associations with personality traits, affective dispositions, and related psychological constructs remains scattered and inconsistent (Brand et al., 2023; Lin et al., 2023; Murphy et al., 2019; Todd et al., 2022). Study 2 therefore had two primary goals. First, to provide a robust overview of the dispositional correlates of the IAS by leveraging the large number of available datasets. Second, to evaluate the practical benefits of using the refined 14-item version of the scale identified in Study 1, compared to the original 21-item unidimensional solution, in order to assess whether structural improvements translate into clearer and more reliable external associations.

Methods

Materials

We selected measures that appeared in multiple datasets from Study 1 and/or were relevant given the scope of the study, i.e., constructs related to physiology, mood, personality, psychopathology, and beliefs and misbeliefs. We merged scores of regular and abridged versions for conciseness where applicable.

Interoception Related. The Multidimensional Assessment of Interoceptive Awareness Version-2 (MAIA-2, Mehling et al., 2018) is a 37-item questionnaire assessing eight dimensions of interoception: Noticing, Not-Distracting, Not-Worrying, Attention Regulation, Emotional Awareness, Self-Regulation, Body Listening, and Trust. Responses are rated on a 6-point Likert scale (e.g., “I notice when I am uncomfortable in my body”).

The Body Perception Questionnaire Short Form (BPQ-SF, Cabrera et al., 2018) contains 46 items measuring Body Awareness and Autonomic Reactivity on a 5-point Likert scale (e.g., “My heart often beats irregularly”). The very-short form (BPQ-VSF) contains 12 items from the Body Awareness subscale. For this study, scores from both versions were combined.

The Interoceptive Confusion Questionnaire (ICQ, Brewer et al., 2016) consists of 20 items rated on a 5-point Likert scale (e.g., “I cannot tell when my muscles are sore or tight”) assessing difficulties in interpreting non-affective physiological states, such as pain or hunger.

The Toronto Alexithymia Scale (TAS-20, Bagby et al., 1994) contains 20 items rated on a 5-point forced-choice scale (e.g., “I have feelings I can’t quite identify”), divided into three dimensions: difficulty identifying feelings, difficulty describing feelings, and externally oriented thinking.

The Bermond–Vorst Alexithymia Questionnaire (BVAQ, Vorst & Bermond, 2001) consists of 40 items across five subscales: fantasising, identifying, analysing, verbalising, and emotionalising. Items are rated on a 5-point Likert scale. The BVAQ also provides two higher-order factors, namely affective and cognitive (e.g., “I like to tell others how I feel”).

Mood and Anxiety Related. The Beck Depression Inventory-II (BDI-II, Beck et al., 1996) is a 21-item measure of depressive symptom severity, including psychological (e.g., guilt) and physiological (e.g., loss of energy) components. Items are rated on a 4-point scale (e.g., “I feel sad much of the time”).

The Patient Health Questionnaire-4 (PHQ-4, Kroenke et al., 2009) is a 4-item screening tool for depression and anxiety, rated on a 4-point Likert scale (e.g., “Little interest or pleasure in doing things”; “Not being able to stop or control worrying”). The PHQ-2 (PHQ-2, Kroenke et al., 2003) and General Anxiety Disorder-2 (GAD-2, Kroenke et al., 2007) scores were combined with the PHQ-4 scores.

The Patient Health Questionnaire-9 (PHQ-9, Kroenke et al., 2001) includes 9 items assessing depressive symptoms on a 4-point Likert scale (e.g., “Feeling tired or having little energy?”).

The Patient Health Questionnaire-15 (PHQ-15, Kroenke et al., 2002) assesses somatic symptom distress with 15 items rated on a 3-point scale (e.g., “Over the last week, how often have you been bothered by back pain?”).

The Short Mood and Feelings Questionnaire (SMFQ, Angold et al., 1995) is a 13-item measure of recent depressive symptoms in children aged 6–17, rated on a 3-point scale (e.g., “I felt miserable or unhappy”).

The State–Trait Anxiety Inventory Trait Version (STAI-T, Spielberger, 1970) is a 20-item trait anxiety questionnaire rated on a 4-point Likert scale (e.g., “I worry too much about something that really doesn’t matter”). Scores from the shorter version of the STAI-T-5 (Zsido et al., 2020) were combined with the STAI-T.

The Generalized Anxiety Disorder Scale (GAD-7, Spitzer et al., 2006) is a 7-item measure of generalized anxiety symptoms rated on a 4-point Likert scale (e.g., “Not being able to stop or control worrying”).

Personality. The NEO Five-Factor Inventory, neuroticism subscale (NEO-FFI, Costa et al., 1992) is a 12-item measure of neuroticism, assessing the general tendency to experience negative emotions such as fear, sadness, or disgust. Items are rated on a 5-point Likert scale (e.g., “I often feel inferior to others”).

The Mini International Personality Item Pool (MINI-IPIP6, Sibley et al., 2011) includes 24 items measuring six broad traits: Agreeableness, Conscientiousness, Neuroticism, Openness, Extraversion, and Honesty-Humility. Items were scored on visual analogue scales in our samples (originally 7-point Likert scales; e.g., “Have a vivid imagination”).

The Big Five Inventory Short Form (BFI-S, Rammstedt & John, 2007) is a 10-item measure of the Big Five traits, rated on a 7-point Likert scale (e.g., “I see myself as someone who ... is considerate and kind to almost everyone”).

Psychopathology. The Personality Inventory for DSM-5 Short Form (PID-5-SF, Thimm et al., 2016) consists of 25

items rated on a 4-point Likert scale, measuring five domains: disinhibition, antagonism, detachment, negative affect, and psychoticism (e.g., “Plenty of people are out to get me”).

The Schizotypal Personality Questionnaire – Brief Revised (SPQ-BR, Davidson et al., 2016) contains 32 items rated on a 5-point Likert scale, assessing four primary dimensions: cognitive-perceptual, interpersonal, disorganized, and social anxiety. These are subdivided into nine secondary factors, including constricted affect, eccentricity, magical thinking, lack of close friends, odd speech, referential thinking, social anxiety, suspiciousness, and unusual perceptions (e.g., “Plenty of people are out to get me”).

The McLean Screening Instrument for Borderline Personality Disorder (MSI-BPD, Zanarini, 2003) is a 10-item dichotomous screening measure for borderline personality disorder (present/absent; e.g., “Have any of your closest relationships been troubled by a lot of arguments or repeated breakups?”).

The Autism Spectrum Quotient Short Form (ASQ-S, Hoekstra et al., 2011) is a 28-item measure of five autistic traits: social skills, adherence to routines, cognitive flexibility, imagination, and patterns/numbers. Items are rated on a 4-point Likert scale (e.g., “I find it difficult to work out people’s intentions”).

Beliefs and Misbeliefs. The Generic Conspiracist Beliefs Scale (GCBS, Brotherton et al., 2013) is a 15-item measure assessing five dimensions of conspiracy beliefs: government malfeasance, extraterrestrial cover-up, malevolent global conspiracies, personal wellbeing, and control of information. Items are scored on a 5-point Likert scale (e.g., “Secret organizations communicate with extraterrestrials, but keep this fact from the public”).

The Primal World Beliefs Inventory (PI-99, Clifton et al., 2019) is a 99-item measure assessing 26 high-order beliefs about the world. Items are scored on a 5-point Likert scale (e.g., “Nearly everything in the world is beautiful”) The PI-18 (Clifton & Yaden, 2021) is a short form of the original questionnaire assessing the 4 main world beliefs: the world is Good (vs. Bad) and its three dimensions: Safe (vs. Dangerous), Enticing (vs. Dull), and Alive (vs. Mechanistic). Scores from the PI-18 scores were combined with the PI-99.

The Lie Scale (LIE, Makowski et al., 2023) is a 16-item measure of lying tendencies across four components: Ability, Frequency, Negativity, and Contextuality. Items are scored on visual analogue scales (e.g., “I am a good liar”).

Data Analysis

All variables from the 16 open-access datasets will be extracted and standardized (z-scored). For continuous predictors, we will first ensure consistent ranges across datasets; when necessary, ranges will be rescaled to achieve comparability. In cases where scales of different lengths measure the same construct, scores will be harmonized: when possi-

ble, means will be computed from overlapping items, while in other cases total scores will be normalized by the number of items. We will collapse redundant subscales (e.g., PHQ-2 and GAD-2 within PHQ-4) into unified composite scores. Similarly, short forms of trait anxiety (e.g., STAI-5-trait) will be collapsed with their longer counterpart (STAI-T), with scores computed as the mean to ensure comparable ranges. Shorter formats of the PI (PI-18) will be combined with scores of its longer version (PI-99). Missing data will be handled via listwise deletion on a per-model basis prior to fitting.

Associations will be estimated using multilevel regression models implemented in the glmmTMB package (McGillycuddy et al., 2025). Each predictor will be entered separately, with the seven IAS item pairs identified in Study 1 (Hungry–Thirsty, Muscles–Pain, Wind–Burp, Urinate–Defecate, Breathing–Heart, Bruise–Blood, Cough–Sneeze) and the original IAS scale serving as outcomes. Models will include random intercepts and slopes by sample to account for between-study heterogeneity; when models fail to converge, a reduced structure with random intercepts only will be used. This approach will treat datasets as clusters, thereby providing pooled estimates of the associations while accounting for both within- and between-sample variability, akin to a meta-analytic framework.

From each fitted model, we will extract standardized regression coefficients (β), 95% confidence intervals, p-values, convergence status, and effective sample size. Participants identifying as “Other” will be excluded from analyses due to limited representation.

Results

Age was positively associated across IAS correlates (mean $r = .13$), with all correlations significant. Gender showed near-zero effects (mean $r = -.06$), reaching significance only with the Hungry–Thirsty, Wind–Burp and Urinate–Defecate pairings.

Several MAIA subscales showed consistent positive associations with the IAS: Noticing, Body Listening, Attention Regulation, Trusting, Emotional Awareness, and Self-Regulation (mean $r \approx .18-.26$), with effects generally stronger for the full IAS, with the exception of the Body Listenig dimension whose associations were stronger with the Bruise–Blood pairing. The Not Distracting subscale was significantly associated with the whole scale and with three of the seven pairs, namely Hungry–Thirsty, Bruise–Blood sugar, and Muscles–Pain (mean $r \approx .19-.13$). By contrast, the Not Worrying subscale was only significant for the full IAS and the Hungry–Thirsty and Urinate–Defecate pairings (mean $r \approx -.009-.108$).

Autonomic Reactivity (BPQ; mean $r = -.19$) was reliably negative across most pairings, while Body Awareness (BPQ; mean $r = .12$) was weak and only reliable for Bruise–Blood.

In terms of interoceptive deficits, alexithymia measures

were consistently negative, with robust effects across both the full IAS and pairings, except for the Affective dimension of the BVAQ, which showed a positive association only with the Bruise–Blood sugar pairing ($r = .16$). Interoceptive confusion assessed with the ICQ was also consistently negatively associated with all pairings and the full IAS (mean $r \approx -.49-.20$).

Depression (BDI; mean $r = -.20$) and anxiety (STAI-T; mean $r = -.19$) were consistently negative across the IAS and most pairings, with Hungry–Thirsty again emerging as a particularly sensitive correlate. Notably, different measures indexing similar constructs showed distinct association patterns: the MFQ was largely nonsignificant despite measuring depressive symptoms comparable to for instance the BDI. Similarly, PHQ-9 followed a weaker and less reliable pattern. Related symptom measures (PHQ-15, PHQ-4, GAD-7) showed weaker but often reliable effects, although some were restricted to specific pairings (e.g., PHQ-15 not significant with Wind–Burp and Breathing–Heart). Notably, most of the mood measures (PHQ-9, BDI, PHQ-4, STAI-T, GAD-7) displayed stronger associations with the Hungry–Thirsty pairing than the full IAS.

Autistic traits were generally negatively associated with the IAS, with Imagination (mean $r = -.18$) showing the most consistent effects across pairings, except for Cough–Sneeze. Across subscales, most associations observed for the full IAS were primarily driven by the first three pairings—Hungry–Thirsty, Bruise–Blood sugar, and Muscles–Pain—which showed the strongest and most reliable negative correlations. Beyond these, effects tended to weaken and often became nonsignificant, particularly for Patterns and Numbers, which did not correlate with any IAS correlate, while Borderline Personality only correlated with the Hungry–Thirsty pairing (mean $r = -.14$).

Schizotypal traits were generally negatively associated with the IAS, with Social Anxiety (mean $r = -.11$) and No Close Friends (mean $r = -.12$) showing the most consistent negative effects across pairings. Notably, although some facets did not show consistent associations with the overall IAS, they were significantly related to specific pairings. For example, Unusual Perceptions was associated with the Muscles–Pain and Urinate–Defecate pairings, whereas Eccentric was associated with the Hungry–Thirsty and Urinate–Defecate pairings.

Within the Big Five, Openness (mean $r = .14$), Agreeableness (mean $r = .05$), and Conscientiousness (mean $r = .03$) showed small but consistent positive associations, most reliably captured by the full IAS but also evident in pairings. Honesty–Humility showed a unique negative association with the full IAS scale, largely driven by the Hunger–Thirst and Bruise–Blood sugar item pairings. Similarly, Neuroticism was negatively associated with the full IAS, but this relationship was primarily attributable to the Hunger–Thirst and Wind–Burp pairings, with the strongest correlation observed

for Hunger–Thirst ($r = -.23$).

Primal world beliefs measures were largely null, but certain beliefs showed small yet consistent associations with specific IAS item pairings. Belief in a comprehensible world (“Understandable”) correlated positively with most pairings except Urinate–Defecate and Cough–Sneeze, while beliefs that the world is “Alive” and “Hierarchical” were associated with the full IAS and Hunger–Thirsty, and “Good” was uniquely linked to Muscles–Pain (mean $r = .09$). For the Lying Profile, only Lying Contextuality showed a notable negative association with the Bruise–Blood sugar pairing ($r = -.12$).

Conspiracy Beliefs showed selective associations, with Global Conspiracies linked to the full IAS and several pairings (Original, Hunger–Thirsty, Cough–Sneeze, Urinate–Defecate; mean $r = .103$), Personal Wellbeing to the full IAS and Hunger–Thirsty (mean $r = .086$), and Information Control to Cough–Sneeze (mean $r = .06$).

Overall, the full IAS yielded the largest number of significant associations (29), but these were often driven by specific item pairings rather than reflecting uniform effects across the scale. The Hungry–Thirsty (17) and Bruise–Blood sugar (10) pairings consistently produced equally strong or stronger effects, particularly for interoceptive, mood, and psychopathology measures. In contrast, other pairings (e.g., Cough–Sneeze, Wind–Burp, Urinate–Defecate) contributed little unique variance and rarely produced meaningful associations.

Discussion

Study 2 examined the dispositional correlates of the IAS and tested whether the refined 14-item version offers practical advantages over the original 21-item scale. The study provides insight into the conceptual boundaries of interoceptive self-reported accuracy across psychological domains.

The IAS converged robustly with other interoceptive self-report measures, particularly facets of the MAIA, and correlated more strongly with dynamic bodily monitoring (Autonomic Reactivity on the BPQ) than general body awareness (Brand et al., 2023; Campos et al., 2021; Todd et al., 2022). Item pairings reflecting visceral sensations, such as Hunger–Thirst, were consistently linked to lower depressive and anxiety symptoms (Brand et al., 2023; Lin et al., 2023), supporting theoretical models connecting interoception to emotion regulation and monitoring of internal bodily states (Khalsa, Adolphs, Cameron, Critchley, Davenport, Feinstein, Feusner, Garfinkel, Lane, Mehling, & others, 2018; Suksasilp & Garfinkel, 2022).

At the same time, these associations were context-dependent: not all item pairings or bodily domains related equally to mood or somatic symptoms (Koike & Nomura, 2023; Murphy et al., 2019). Emerging evidence suggests that interoceptive abilities (e.g., sensitivity, precision and

metacognitive insight) may not uniformly predict mental health outcomes and may depend on measurement approach or sample characteristics (@ Banellis et al., 2025).

Associations with broader personality traits and high-order beliefs were modest. Small positive correlations with Openness and Agreeableness and weak associations with primal world beliefs (e.g., perceiving the world as alive or comprehensible) suggest that interoception may subtly interact with higher-level cognitive frameworks (Gabriele et al., 2022; Hausteine et al., 2024; Simkute et al., 2025). However, the effects were weak and inconsistent, reinforcing that interoception primarily aligns with bodily self-perception rather than broad personality or worldview constructs.

Certain item pairings (e.g., Hunger–Thirst, Bruise–Blood sugar) captured associations with mood and interoception as strongly as, or stronger than, the full 21-item scale. This highlights the contextual nature of interoceptive accuracy: interoceptive tendencies manifest variably across bodily domains (Desmedt et al., 2022; Khalsa, Adolphs, Cameron, Critchley, Davenport, Feinstein, Feusner, Garfinkel, Lane, Mehling, Meuret, et al., 2018), emphasizing the need for context-specific items.

The refined 14-item IAS performed comparably to the original scale, and item pairings sometimes provided greater sensitivity to specific interoceptive or affective domains, demonstrating that structural refinement can enhance efficiency and interpretability without sacrificing validity.

Study 2 shows that interoceptive self-report is adaptive, context-sensitive, and most predictive in relation to mood and dynamic bodily monitoring. The refined IAS preserves construct validity while enabling more targeted assessment of domain-specific interoceptive tendencies.

General Discussion

The present study provides a comprehensive evaluation of the IAS, examining both its structure (study 1) and its dispositional correlates (study 2). These studies offer novel insights into the psychometric characteristics, conceptual interpretation, and practical applications of the IAS. In this discussion, these findings are contextualized within the broader landscape of interoception research.

Study 1 demonstrated that the IAS is best conceptualised as comprising tightly coupled item pairs, reflecting discrete bodily domains such as visceral sensations, musculoskeletal cues, and expulsion phenomena. The pairs emerged after several items (e.g., Tickle, Taste, Affective Touch) were removed due to their instability, redundancy, or poor associations with the other items. The resulting 14-item IAS retained the psychometric integrity of the original scale while improving interpretability and efficiency.

A short version of the IAS has been proposed before. Lin et al. (2023) suggested a 12-item version by removing items with local dependency, high difficulty, and/or low discrimi-

nation. Their parallel analysis suggested unidimensionality, although parallel analysis is known to underestimate factor numbers in multidimensional data (H. F. Golino & Epskamp, 2017; Markos & Tsigilis, 2024).

In contrast, our study applied multiple methods including EGA, FA, and HCA yet the same pairs consistently emerged across datasets and demonstrated superior fit compared to alternative models, which challenges the assumption of unidimensionality of the IAS (Brand et al., 2023; Murphy et al., 2019). Although the short version of the IAS by Lin et al. (2023) favoured a one-factor solution, their 12-item IAS was derived by removing locally dependent items identified through Item Response Theory (IRT).

We found that these locally dependent items can instead be interpreted as meaningful pairs (e.g., Heart–Breath), reflecting distinct bodily domains. Because IRT assumes unidimensionality (Nguyen et al., 2014), it may have treated these natural item clusters as statistical artefacts rather than as theoretically relevant structures within interoceptive accuracy.

Study 2 provided further evidence for a domain-specific perspective. Certain item pairings, particularly Hunger–Thirst, Bruise–Blood sugar, and Muscles–Pain, showed consistent negative correlations with depression and anxiety measures, sometimes approaching or even surpassing the strength of associations observed for the full IAS scale. Notably, Hunger–Thirst and Bruise–Blood sugar were also the only pairs, out of the seven examined, that displayed significant positive correlations with the BPQ Body Awareness subscale, whereas the full IAS scale showed no such association.

Additionally, specific item pairings revealed distinct patterns of association with other interoceptive measures and mental health indices that were not captured by the full IAS. For instance, while the full IAS showed no significant associations with Borderline Personality, the Hunger–Thirst pair exhibited negative correlations. These findings collectively suggest that item-level analyses may uncover more nuanced relationships than those identified by the overall scale. This pattern highlights that questionnaires like the IAS may capture multiple facets of interoception rather than a single underlying dimension. Consequently, adopting a domain-specific approach could enhance construct validity by better reflecting the heterogeneity of interoceptive experiences.

These findings are in line with inconsistencies in the literature concerning associations between the IAS and the BPQ Body Awareness. While some studies report significant negative associations (Lin et al., 2023), others have observed significant positive associations (Brand et al., 2023; Campos et al., 2021; Gaggero et al., 2021; Todd et al., 2022), and still others none at all (Murphy et al., 2019). In our study the full IAS was also not associated with the BPQ Body Awareness.

This pattern is consistent with the idea that this subscale may capture maladaptive bodily experiences or autonomic dysfunction, rather than interoceptive accuracy per se. How-

ever, the Body Awareness subscale of the BPQ has also been criticized for its ambiguous instructions. It is often unclear whether respondents should interpret the items as assessing attentional focus on bodily sensations, perceived accuracy in detecting them, or simply the frequency of experiencing such signals. As a result, studies have reported that observed associations vary depending on how the BPQ Body Awareness is interpreted (Campos et al., 2021; Gabriele et al., 2022; Gaggero et al., 2021). As such, it is unclear whether the lack of a relationship between the IAS and the BPQ Body Awareness in this study reflects a genuine divergence between the constructs measured by the two scales, or instead arises from ambiguities in how the BPQ Body Awareness is operationalized and interpreted across studies.

We highlight that some IAS pairings may capture interoceptive self-reported accuracy more effectively than the full IAS measure, raising questions about the interpretability of the full scale. In contrast, other pairings (e.g., Cough–Sneeze, Wind–Burp) were weakly or inconsistently related to other variables, raising concerns about their usefulness. Together, these findings indicate that the strength and relevance of interoceptive self-reported accuracy may vary systematically across physiological domains.

At the same time, the interpretation of such associations is complicated by the lack of contextualisation in the items themselves. For example, a racing heart is often associated with anxiety, yet the same sensation can also arise during exercise or other forms of physiological arousal. Without contextual anchors, it remains unclear which bodily states participants are drawing upon when responding, highlighting the interpretive ambiguity of self-reports. In line with prior work emphasizing the importance of contextual factors in the measurement of interoception (Vlemincx et al., 2023), our results suggest that the value of self-report lies less in broad aggregation across domains and more in the targeted assessment of physiologically and situationally meaningful contexts.

The dispositional correlates of the IAS help delineate the conceptual boundaries of interoceptive accuracy as a construct. Self-reported interoceptive accuracy, as assessed by the IAS, has been robustly associated with mood-related outcomes, particularly depression and anxiety, underscoring its relevance for affective regulation and bodily state monitoring (Benau, 2023; Brand et al., 2023). This aligns with a broader body of research linking affective conditions such as anxiety and depression with altered interoceptive processing (e.g., Garfinkel et al., 2015).

However, recent findings have also reported null or inconsistent associations between interoceptive abilities and mental health symptomatology, suggesting that such relationships may not generalize across all interoceptive domains or measurement modalities (Banellis et al. (2025); Jenkinson et al. (2024)). For instance, a recent review by Jenkinson et al. (2024) found no consistent evidence linking cardiac in-

teroceptive accuracy tasks with anxiety or depression across multiple meta-analyses. Collectively, these findings indicate that relationships between interoceptive accuracy and mental health may vary across bodily domains and measurement modalities. This suggests that the IAS, while informative, may benefit from a more differentiated approach that accounts for domain-specific and contextual influences on self-reported interoceptive accuracy.

In contrast to the relatively stronger associations observed with affective measures, correlations between IAS scores and personality traits were small but directionally consistent. Modest positive associations emerged with Openness, Agreeableness, and Conscientiousness, while Neuroticism and Extraversion showed near-zero relationships, and Honesty–Humility displayed weak negative correlations. This pattern suggests that self-reported interoceptive accuracy, as indexed by the IAS, shares limited but systematic overlap with broad dispositional tendencies—particularly those reflecting intellectual curiosity and creativity, patience and tolerance in interpersonal interactions, and diligence and organizational tendencies (Ashton & Lee, 2007).

Recent evidence supports the view that relationships between interoceptive measures and personality are multifaceted rather than uniform. For instance, Haustein et al. (2024) reported significant associations between interoceptive awareness (MAIA) and all personality traits except Agreeableness, with Openness showing positive correlations with Body Listening and Emotional Awareness, and Neuroticism showing negative correlations with Self- and Attention Regulation. Similarly, Simkute et al. (2025) found that task-based indices of interoceptive sensitivity related negatively to Extraversion. Collectively, these findings indicate that interoceptive constructs show distinct and sometimes opposing patterns of association across measurement modalities, suggesting that the IAS relates to a specific, self-evaluative aspect of interoceptive functioning rather than reflecting a general interoceptive or personality-based trait.

A potential key reason for discrepancies in observed associations, including those seen with the IAS, is that self-report questionnaires in interoception often tap into distinct constructs rather than a common underlying dimension (Desmedt et al., 2022). Measures labeled as interoceptive sensibility, awareness, or self-reported accuracy capture partially overlapping but empirically distinct facets, with low intercorrelations between questionnaires and items. Consequently, differences in associations with psychological outcomes or personality traits may reflect differences in what each questionnaire assesses rather than true variation in interoceptive functioning.

These two studies revealed three main points. First, in the datasets that included analogue-style ratings, these scales captured greater variability, further enhancing the ability to detect individual differences. Second, a shortened, 14-item

version of the IAS retained strong psychometric integrity while removing unstable or redundant items. Third, specific item pairings allowed more precise differentiation among individuals and often outperformed global scores in predictive value. Nevertheless, even this refined IAS has important limitations.

While the 14-item version improves interpretability and efficiency, it continues to underrepresent key interoceptive modalities such as cardiac, respiratory, and nuanced somatosensory signals, limiting its ability to capture the full breadth of interoceptive processes. This shortcoming underscores the need for broader coverage in future scale development. At the same time, analogue-style items and item pairings offer clear advantages: they increase response variability, allow more precise differentiation among individuals, and provide domain-specific indicators that often outperform global scores.

Building on these insights, a next-generation interoception scale could balance breadth (multiple items per modality) with depth (domain-specific subscales), retain the benefits of analogue or wider rating ranges (e.g., 1–7 instead of 1–5), and optimize coverage across physiologically and psychologically relevant bodily domains. Such a scale would improve interpretability, sensitivity, and practical utility, ultimately enabling a more nuanced mapping of interoceptive abilities onto affective, cognitive, and health-related outcomes.

One limitation worth noting is that we did not evaluate the IAS alongside other widely used questionnaires, such as the BPQ or MAIA. Although both are frequently employed in interoception research, they present important limitations. For example, the MAIA-2 contains subscales (e.g., Not-Worrying, Not-Distracting) with weak reliability and minimal relation to a general interoception factor (Ferentzi et al., 2021; Rogowska et al., 2023). Several other subscales primarily reflect higher-level cognitive tendencies, such as mindfulness and attentional regulation, rather than interoceptive body awareness (Todd et al., 2022; Vig et al., 2023). Similarly, the BPQ's Body Awareness and Autonomic Reactivity subscales show conceptual and psychometric concerns, and BPQ scores often overlap substantially with negative affect, raising questions about their specificity as measures of interoceptive sensibility (Campos et al., 2021; Gabriele et al., 2022; Vig et al., 2023).

Because the IAS is a newly developed questionnaire and the only one designed to assess interoceptive self-reported accuracy - a dimension previously assessed solely through objective tasks such as the HDT and HCT - our focus was on evaluating its structure and correlates without introducing established measures with documented shortcomings. Nonetheless, future research should employ novel methods, such as Exploratory Graph Analysis, to clarify the convergent and divergent validity of these questionnaires across interoceptive domains.

Taken together, these studies advance understanding of the IAS and interoceptive self-report more broadly. The refined 14-item IAS provides a more efficient and interpretable measure, capturing domain-specific interoceptive signals that show differential associations with psychological outcomes. Associations were consistently stronger with mood measures and compared to personality and beliefs, indicating that self-reported interoceptive accuracy relates more closely to affective experience than to stable dispositional traits.

At the same time, current findings highlight limitations in modality coverage and item specificity, suggesting that future scale development should incorporate multimodal, context-sensitive, and continuous analogue ratings or wider discrete scales to better capture the richness of interoceptive processes and clarify their role in affective, cognitive, and health-related outcomes (Desmedt et al., 2022; Vlemincx et al., 2023). Ultimately, these studies show that the IAS is most informative when treated as a set of domain-specific, context-sensitive indicators rather than as a broad global index, advancing both the measurement and conceptual understanding of interoception.

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