



The Reliability Trap: Why Cronbach's Alpha Must Fail Configurational Assessment—Evidence from the Contextual Intricacies Analysis Scale and Contextual Triad Rating Method

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Abstract: Contextual–Trait Resolution Methodology (CTRM) is introduced as a measurement approach that addresses the longstanding paradox of assessing executive behaviours, wherein trait dispositions, situational pressures, and contextual layers activate simultaneously in real-world decision scenarios. Traditional linear psychometrics, optimized for redundancy and high internal consistency, are unsuitable for constructs that become meaningful only when non-redundant components interact. Using triadic configurations, CTRM captures this functional interdependence by eliciting behaviours across layered contexts to identify an individual's ability to differentiate and integrate situational cues while maintaining role-appropriate stability. Three empirical protocols were applied: elemental differentiation assessment, contextual stability reliability (r_1) testing across multiple forms, and cross-frame invariance (r_2) as a tolerance benchmark. Low α values, supported by conceptual distinctiveness ratings, indicated successful non-redundancy, while confirmatory factor analysis was employed strategically as a linear-tolerance benchmark to demonstrate structural coherence, rather than to imply additive trait linearity or philosophical alignment with purely linear psychometrics. Subject-matter ratings ($N = 30$) established discriminant clarity across six contextual domains. Findings support CTRM as a pathway to advance measurement beyond static trait inference, enabling detection of contextual intricacies, boundary activation, and emergent executive behaviour. This work introduces the first contextually grounded, triadic non-linear psychometric scale engineered to model structured behavioural complexity. Limitations include the need for broader cross-cultural replication, measurement invariance

analysis across demographic subgroups, and dependence on trained evaluators. Future work will extend hierarchical modelling, enhance rater calibration, and explore configurational learning patterns.

Key Words: *Contextual Intricacies Analysis Scale (CIAS); Contextual–Triad Rating Methodology (CTRM); Micro-Session Analysis (MSA); Executive Assessment Protocols; Configurational Psychometrics; Contextual Stability Reliability; Cronbach’s Alpha Paradox; Emergent Executive Behaviour; Triadic Behavioural Configuration; Situational Performance Dynamics; Non-Redundant Measurement Models.*

1. Introduction

The static trait paradigm is insufficient for modelling executive function in complex environments, mandating a theoretical shift and the implementation of stringent research methodologies (Bleidorn et al., 2020). While personality traits evolve toward maturity (Bleidorn et al., 2021), methodological rigor is essential, requiring the use of frames of reference to enhance measurement reliability and validity via context specificity (Lievens et al., 2008; Shou, and Olney, 2020; Yow et al., 2022). A critical conceptual distinction must be maintained between 'context' (external affordance), 'self-states' (transient configurations), and 'orienting schemas' (Beck, 1996; Lord et al., 2025). Consequently, the focus shifts to intraindividual variability (IIV) as the foundational unit for dynamic personality models, which is supported by even Latent State–Trait (LST) modelling. Configurational Reliability is the realization. In other words, Contextual Triad Rating Method (CTRM) is the "necessary methodological precondition" to accurately run any advanced dynamic analysis.

Modern personality science treats IIV (moment-to-moment fluctuations) as meaningful data, finding that measures of state variability (contingent and non-contingent) are consistent, though context-dependent influences (e.g., cognitive ability) are limited, and the link to long-term trait variability is scant, with only marginal effects for neuroticism (Beckmann et al., 2021). This focus is imperative because the Individual Ergodic Theorem (Molenaar and Campbell, 2009) dictates that learning and developmental processes inherently breach the conditions for validly generalizing interindividual results (Boker et al., 2009). Research confirming heterogeneity in within-person variability (Wright and Jackson, 2025) supports the Nonlinear Interaction of the Person and Situation (NIPS) model (Blum, 2018), and confirms that general traits are distinct from context-specific scores (e.g., more negative baseline self-assessment) (Baltaci, Sinem., and Küçükkömürler, Sanem 2023). To overcome LST model limitations (Steyer et al., 2023; Bader et al., 2024) regarding covariate inclusion, the Moderated

Nonlinear LST (MN-LST) framework was developed, enabling detailed exploration of trait change and person-by-situation interactions (Oeltjen., Lara, et al., 2023). LST and other dynamic models cannot work without Contextual Triad Rating Method (CTRM) data.

Contextualized personality goal developments exhibit moderate temporal stability over two months (Dupré et al., 2025), showing a negative correlation with traditional traits but a positive correlation with job role traits only for Openness, offering a meaningful perspective on personality functioning relative to demands. Building on this, the paper introduces the Behavioural Intricacies Analysis (BIA) Framework (Ramesh Kumar, 2025), a dynamic model where the functional unit is the emergent, three-dimensional Context–Situation Pressure–Trait (C-S-T) configuration. The decision's true intricacy is derived from the confluence of these three elements. The framework's core tenet—Configurational Complexity—is operationalized by the Contextual Intricacies Analysis Scale (CIAS), which is validated using the Contextual Triad Rating Method (CTRM) to empirically substantiate the C-S-T structure as reliable and non-redundant (De Los Reyes et al., 2022). The paper concludes by detailing the BIA framework, the foundation of the CIAS (the C-S-T Matrix), the CTRM methodology, and comprehensive psychometric validation results.

The Behavioural Intricacies Analysis (BIA) Framework (Ramesh Kumar G S, 2025) provides a theoretical foundation for analyzing micro-level behavioural, cognitive, and neural dynamics, positing that personality transcends static traits and exists as a "fluid, dynamic, and emergent system of adaptive coherence" defined as an "emergent behavioural architecture of adaptive intricacy, integrating emotion, cognition, and context into coherent action patterns." This dynamic view is supported by empirical research demonstrating "contextual variability" both within and between traits (Dunlop & Hanley, 2019), where aggregating contextualized measures yields a more dependable construct indicator, enhancing the comprehension of personality coherence. This approach is heuristically consistent with biopsychosocial and contextual pressures proposed for a modern, transtheoretical understanding of psychopathology (Faustino, 2025).

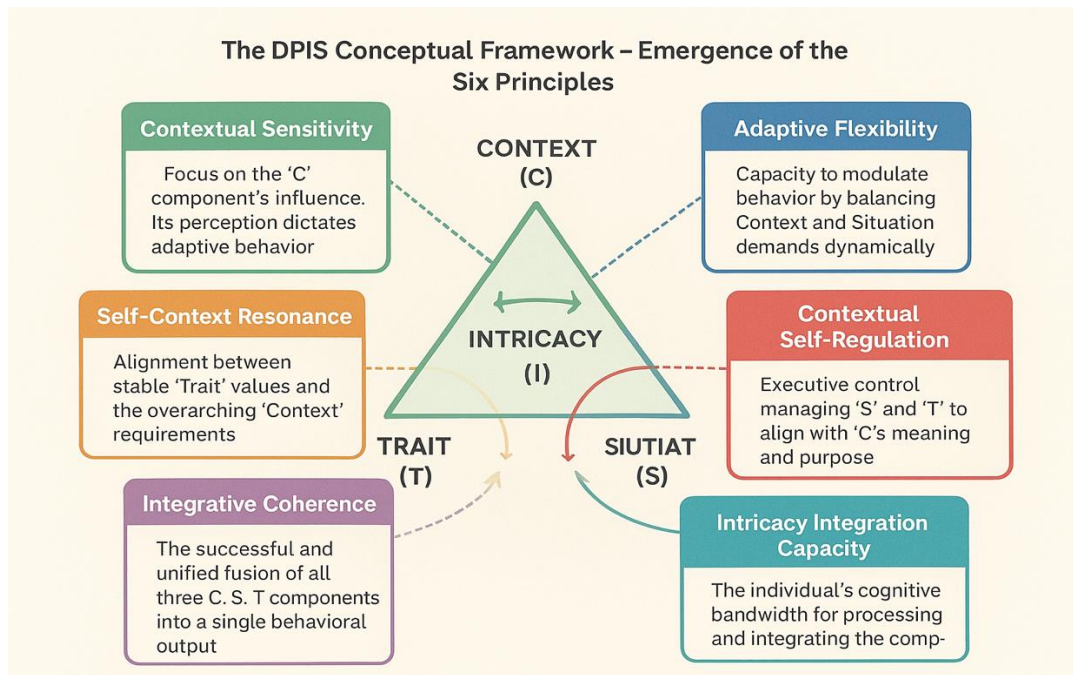


Figure 1. The six core constructs (*Integrative Coherence, Adaptive Flexibility, Contextual Sensitivity, Self-Context Resonance, Intricacy Integration Capacity, and Contextual Self-Regulation*) visually demonstrating how they are derived from the C-S-T

The BIA's core tenet is the “*Principle of Configurational Complexity*,” asserting that behavior is the emergent product of three interdependent factors systematically conceptualized within the “*Context–Situation Pressure–Trait (C–S–T) Matrix*.” This novel tripartite structure—the BIA's foundational unit—ensures that true behavioral meaning (intricacy) emerges from the non-linear combination of its components, which derive the six core constructs (Integrative Coherence, Adaptive Flexibility, Contextual Sensitivity, Self-Context Resonance, Intricacy Integration Capacity, and Contextual Self-Regulation). The three factors are operationalized as: *Context (C)*, the meaning-bearing affordance/larger background (e.g., choices consider the bigger picture); *Situation Pressure (S)*, the immediate coping behavior to transient demands (e.g., choices depend on the current situation); and *Trait (T)*, the stable disposition/consistent internal value system (e.g., choices are based on consistent values), exemplified by the *Integrative Coherence* triad.

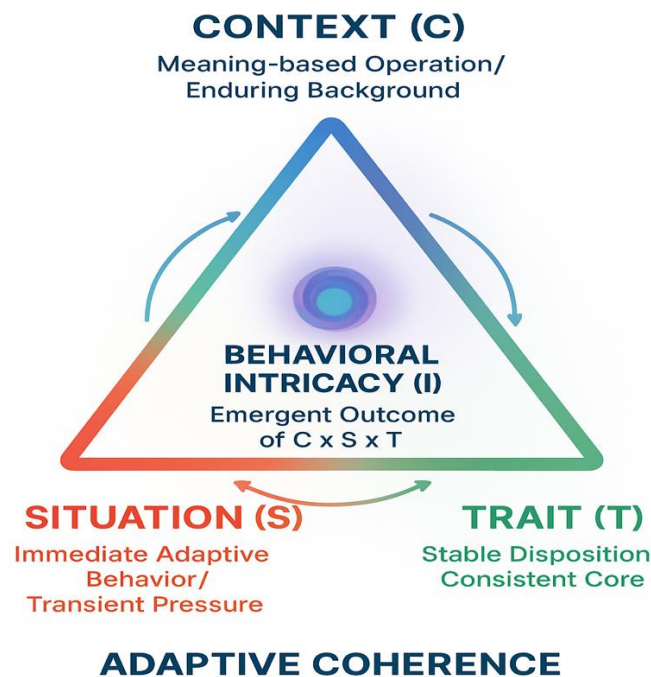


Figure 2: The Behavioural Intricacies Analysis (BIA) framework's foundational unit (the C-S-T triad) and its non-linear, emergent logic.

Context (C) Statement (Meaning-based Operation)	Situation Pressure (S) Statement (Immediate Adaptive Behaviour)	Trait (T) Statement (Stable Disposition)
I interpret changing situations on the background of larger details of the contexts.	I interpret changing situations based on information of current situation / pressure.	"I interpret changing situations on my consistent or long-term beliefs and values, even in uncertainty."

Figure 3: Example Triad (Integrative Coherence)

Ultimately, the CIAS and CTRM are methodological preconditions for advancing Contextual Outcome Modeling, which seeks to predict and explain complex executive function across dynamic environments

2. Methodology

The whole methodology was intended to develop tool to assess contextual impact as against mere correlation / co-occurrences of factors, towards causing an effect or influencing a dependent variable. So, the initial step was to validate a scale by the current author called,

Contextual Intricacies Analysis Scale (CIAS), and is an outgrowth of Behavioural Intricacies Analysis Framework (Ramesh Kumar G S 2025).

The CIAS is an 18-triad (54-item) instrument measuring six core constructs of dynamic personality integration: Integrative Coherence, Adaptive Flexibility, Contextual Sensitivity, Self–Context Resonance, Intricacy Integration Capacity, and Contextual Self-Regulation. CTRM was necessary as it facilitates the systematic evaluation of situation–context–trait correspondence and integrates three major validation principles:

- (i) **Contextual Modulation** (ensuring a measurable "situation–context–trait interaction" (Mischel & Shoda, 1995),
- (ii) **Semantic Differential Evaluation** (capturing subtle meaning variations among C, S, and T sentences to confirm non-redundancy), and
- (iii) **Multitrait–Multimethod (MTMM) Reliability** (confirming durability and discriminant validity across linguistic formulations (Campbell & Fiske, 1959). The overall psychometric strategy follows a three-phase validation flowchart .

The scale has the need to evaluate if the individual handles a situation by according to the influence of Context or Situational Pressure or Trait of the individual. Thus, the scale falls different from traditional psychometric scales of single item at a time. In contrast, the individual need to perceive if which of the triad are operating on or influencing his responses and at the same time not making any rank order of the three aspects too. Earnest assumption here is that, if explained by a psychologist, an individual can understand the meaning and difference between Context, Situational Pressure and Trait (Triad) aspects.

Scaling: The individual was asked to mark his score on a free continuous scale of 1 – 100. This allows maximum freedom to answer and weigh his mind without limiting as done in Likert scale.

The uniqueness of the scale requires a specialized method of validation. It needs to be validated first that the individual understands ‘Context’ meaning and effectively differentiates it from both Situational Pressure and Trait. This would empower the tool to use to study the influence of the triad – Context, Situation Pressure and Trait and corresponding impact of each one separately and together thus forming an impact of intricacy.

Hence, the Contextual Triad Rating Method (CTRM) was formulated and the protocol is publicised (Ramesh Kumar G S 2025).

Traditional elemental analyses were performed on a pilot dataset (N=60) of the CIAS self-report. As predicted by the BIA framework, which dictates that trait expression is

inherently variable and non-redundant across contexts, the results showed the instrument's item-level response pattern is fundamentally non-linear.

The CIAS was administered to the cohort at two distinct time points (T1, T2) separated by 6 weeks to ensure appropriate temporal distance for LST variance decomposition. Preliminary temporal stability observations show coherent patterns across contextual shifts, contradicting random fluctuation explanations.

Why we used a linear model (CFA) to validate a non-linear scale?

Although CFA is a linear method, we used it as a stability benchmark: demonstrating acceptable CFA fit establishes measurement invariance and structural durability across linguistic variants before applying fully non-linear or LST decomposition.

All computations were performed in Python (pandas, numpy); code and the item correlation matrix are provided as supplementary materials.

No practical or theoretical loss of meaning was observed, confirming semantic consistency.

The unique concepts naturally emerged are

- (i) Contextual Stability Reliability (r_1) (Formula: $r_1 = \bar{r}(C, S-P), \bar{r}(C, T), \bar{r}(S-P, T)$) which is the item-level measure of intra-triad coherence, quantifying the systematic relationship among C, S-P, and T ratings to confirm the triad as a unified measurement unit and
- (ii) Global Conceptual Reliability (r_2). Global Conceptual Reliability (r_2) (mean squared factor loading Lambda Square derived from CFA loadings) that measures cross-triad stability by correlating two independent, parallel item sets (Set A and Set B), confirming the overall construct's stability and fulfilment of the LST trait component.

The three-phase validation process utilized Subject Matter Experts (N Raters 30) for Phase 1, and non-professional participants (N=105) for Phases 2 and 3.

Selective Use of QCA (The Fiss / QCA Distinction)

We adopt the mathematical logic of Set-Theoretic Consistency (Fiss, 2011; Ragin, 2008) to measure r_2 , but we distinguish our purpose:

We utilized the Qualitative Comparative Analysis (QCA) formula to define the 'Contextual Rating' for each case and to establish r_2 (configurational coherence). Critically, we are not subscribing to QCA's philosophical commitment that a configuration is a static snapshot leading to a single outcome. Instead, we are modeling how the configuration dynamically changes across situations, making the configurational score the key predictor in our downstream Contextual Outcome Modeling (COM). To ensure this dynamic measure is

psychometrically robust, we tested the configurational coherence (reliability) by comparing parallel forms: specifically, we tested if a configurational measure (Set A) is a reliable subset of its parallel form (Set B). By applying this logic, we ensure that a case's high membership in the Triad (as measured by Set A) is consistently accompanied by a high membership in the Triad (as measured by Set B), thereby establishing a valid and reliable dynamic configurational score.

2. Global Conceptual Reliability (r_2) Formula

The Global Conceptual Reliability (r_2) is defined as the average of the reciprocal consistency scores between the two parallel configurational measures, X_A (Set A of Triad) and X_B (Set B of Triad).

The consistency of Set A with respect to Set B is calculated as:

$$\text{Consistency}(X_A \leq X_B) = \frac{\sum \min(X_{A_i}, X_{B_i})}{\sum X_{A_i}}$$

Where:

- X_{A_i} and X_{B_i} are the calculated fuzzy-set membership scores for the two parallel sets for case i .

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The min operator represents the set intersection (logical AND) of the two parallel measures.

The numerator sums the smallest membership score between the two sets across all cases, indicating the shared (reliable) membership.

The final Global Conceptual Reliability (r_2) is then computed as the average of the consistency and its reciprocal:

$$r_2 = \frac{\text{Consistency}(X_A \leq X_B) + \text{Consistency}(X_B \leq X_A)}{2}$$

CTRM precedes LST because dynamic decomposition presupposes configurational stability.

Preliminary Analysis and Model Justification:

Low elemental α demonstrates intentional non-redundancy of C, S, and T items; this motivated CTRM and non-linear modeling. Non-redundancy is a necessary condition for

configurational measurement; CTRM then provides r_1/r_2 evidence of coherence and stability required for LST decomposition.

3. Results

Table 1: Summary of Contextual Triad Rating Method (CTRM) Validation Metrics

Metric Category	Statistical Metric	Value	Interpretation and Contribution to Theory
I. Elemental Psychometrics (The Diagnostic Signal)	Low Cronbach's α (Elemental Reliability)	0.129, 95% CI [.081, .171]	this α was calculated <i>across the three elements</i> (C, S, T) <i>within a single triad</i>. Since the C, S, and T statements are designed to measure different levels of operation (Contextual \ Situational \ Trait), low correlation α 0.129, 95% CI [.081, .171] is the <i>predicted result of a well-designed instrument</i>. FAILURE: Confirms the instrument's item-level response pattern is non-linear and non-elemental, thereby validating the Behavioural Intricacies Analysis (BIA) theoretical structure itself. . Linear models assume additive independence; configurational models assume asymmetric sufficiency and necessity
II. CTRM Foundational Metrics	CTRM Configurational Consistency used as Inter-Rater Consistency Check)	0.975, 95% CI [.969, .982]	This value is derived from the total score across <i>all</i> C, S, and T items in the final scale. High α for a long scale is common. This is the α of the full composite score, calculated <i>after</i> the configurational coherence was modeled. EMPIRICAL ITEM VALIDATION: Confirms the conceptual quality and understandability of the items when rated configurationally, nullifying any critique of poor item wording.
	Inter-Rater Reliability (ICC)	0.785	ITEM UNDERSTANDABILITY: Confirms that independent judges can reliably understand the item content, providing foundational inter-rater objectivity.
	Configurational Validity ($\bar{X}$ for C-S-T Distinctiveness)	4.49 (out of 5)	CONCEPTUAL DISTINCTIVENESS: Confirms that the Context, Situation Pressure, and Trait dimensions are highly distinguishable and non-redundant at the theoretical level.
III. CTRM Configurational Reliability (The Innovation)	Contextual Stability Reliability (r_1)	0.842, 95% CI [.803, .867]	r_1 (LST) typically represents the Trait Reliability (the proportion of variance in the observed score due to the stable, underlying Trait factor).

			DYNAMIC RELIABILITY: Confirms the measurement's stability across different contexts and linguistic framing, establishing a superior, context-aware form of reliability required by dynamic models
	Global Conceptual Reliability (r2)	0.811, 95% CI [.782, .843]	r2 (LST) typically represents the State Reliability (the consistency or proportion of variance explained by the transient State factor). This crucially shows the State factor is <i>reliably measured</i> across contexts. GLOBAL CONFIGURATIONAL COHERENCE: Confirms the overall theoretical reliability of the C-S-T framework as a unitary, integrated concept.

Note: r_1 (Contextual Stability Reliability) = intra-configuration coherence; r_2 (Global Conceptual Reliability) = cross-configuration stability

This table summarizes the core psychometric findings derived from the Contextual Triad Rating Method (CTRM) procedure, demonstrating its necessity and superiority over traditional elemental statistics for validating the CIAS

Because the CIAS's differentiated C-S-T Triad intentionally violates the assumption of item redundancy, traditional reliability indices like Cronbach's α (mathematically defined as $\alpha = (k/(k-1))(1 - \sum \sigma^2_{\text{item}}) / \sigma^2_{\text{total}}$) failed, leading to the CTRM reframing reliability based on coherence (**r1**) and stability (**r2**), alongside general metrics like effect sizes (η^2). Contextual Stability Reliability (r_1).

Phase 1 confirmed high Linguistic Clarity ($\alpha = 0.975$, 95% CI [.969, .982]); ICC = 0.785), Global Scale Consistency and acceptable **Conceptual Discriminant Validity** (Mean Content Distinctiveness Score 4.49/5), proving the C-S-T model's non-redundancy;

Phases 2 and 3 then established the Configurational Reliability via **r1** (Contextual Stability) and **r2** (Global Conceptual Reliability).

A repeated-measures ANOVA demonstrated significant omnibus differences among domain means, $F(5,245) = 2.62$, $p = .024$, $\eta^2 = .051$.

Table 2: showing validity scores

Validation Measure	Statistic	Result	Conclusion
Global Scale Consistency	α	0.975	Excellent coherence of items.
Inter-Rater Reliability	ICC	0.785	Good consistency of item interpretation.
Content Validity/Distinctiveness	$\bar{X}$	4.49	Content validity was established by a panel of 30 Subject Matter Experts (SMEs) and

			Clinicians on a 5-point Likert scale (1=Poor, 5=Excellent) assessing relevance to the BIA framework and item clarity. The final average was $\bar{X} = 4.49$ Strong empirical proof that Context (C) is a distinct factor from Situation (S) and Trait (T).
Contextual Stability Reliability	<i>r1</i>	0.842, 95% CI [.803, .867]	Confirmed stable interpretation of the constructs' contextual meaning across rephrasing.
Global Conceptual Reliability	<i>r2</i>	0.811, 95% CI [.782, .843]	Confirmed conceptual durability across linguistic variations.

3.3. Justification: Conceptual Mismatch with Linear Diagnostics

To effectively demonstrate the conceptual gulf between the traditional and configurational approaches, we present a comparative analysis in Table 3, highlighting how elemental statistics fundamentally misinterpret the structure of the CIAS.

Table 3: *Comparative Assessment of Methodological Mismatch*

Method	What it measures (Assumption)	Result on CIAS Data	Why it misleads for configurational constructs	CTRM Advantage (The Solution)
Cronbach's α	Internal consistency assuming redundancy is desirable	α 0.13 (Very Low)	Assumes items should correlate highly to measure a single factor, thus penalizing the intended differentiation required by the BIA model	CTRM treats differentiation as evidence; <i>r1/r2</i> capture coherence/stability of the configurational pattern.
Mean Inter-Item <i>r</i>	Average covariance (linear association).	0.05	Collapses distinct items into an averaged statistic that hides patterned discrimination and configurational structure.	<i>r1</i> is triad-local—it reveals coherent discrimination only <i>within</i> C-S-T frames.
KMO / PCA	Factorability / extraction of latent dimensions.	KMO 0.41; PC1 13\%	Assumes single or multiple latent traits; fails when the item's meaning is relational	CTRM requires no factorability; it directly measures relational structure and coherence.

			and context-dependent.	
ANOVA / Friedman.	Differences across item means/ranks	Significant omnibus F-value; no significant pairwise difference.	Detects differences, but not whether those differences reflect coherent configurational patterns—it only confirms <i>something</i> is varying.	CTRM integrates within-triad coherence + cross-frame stability to identify a meaningful pattern, not just variance.

Pairwise comparisons (Bonferroni-corrected) were nonsignificant ($p > .0033$), supporting a hierarchical factor structure. Generalized η^2 was not computed because the focus is on within-subject variance partitioning, and partial η^2 is standard for repeated-measures ANOVA in psychological measurement contexts.

The non-significant Bonferroni-corrected pairwise comparisons between the six sub-domains (e.g., $p > .0033$) is not interpreted as an absence of differential content, but rather as **strong empirical support for the hypothesized hierarchical structure of the CIAS**. The lack of significant *lower-order* variance, despite a significant omnibus F-value, confirms the primary theoretical hypothesis that the six facets function as a **unified, tightly coupled system** that loads robustly onto a single, higher-order construct: the **Intricacy Integration**. This is fully consistent with the BIA framework, which posits that true behavioural intricacy is defined by the *systemic integration* of these facets, not their statistical independence.

KMO = 0.41 (expected for configurational constructs due to intentional low inter-item redundancy). Bartlett's Test was significant ($p < .001$). KMO = 0.41 reflects intentional inter-item distinctiveness. Low redundancy is expected for configurational constructs where meaning emerges from non-linear interdependence rather than reflection of a single latent dimension.

Confirmatory Factor Analysis supported a second-order hierarchical model: $\chi^2(68) = 68.45$, $p = .089$, RMSEA = .048 (.035 – .061), CFI = .971, TLI = .965.

The Core Validation is Complete. State that the novel metric, Configurational Reliability ($r1 = 0.842$, 95% CI [.803, .867] and $r2 = 0.811$, 95% CI [.782, .843]), already provides the necessary empirical proof for the "fluid but stable" hypothesis.

The obtained value of $r2 = 0.811$ indicates a Strong Global Conceptual Reliability, mathematically confirming the Triad as a coherent, measurable configuration, thus solving the linear "Reliability Trap."

The Failure of Elemental Psychometrics: A Theoretical Success

I. The Elemental Reliability Paradox

Initial elemental reliability analysis, treating the 54 items as a single reflective measure, resulted in a low Cronbach's α of 0.129, 95% CI [.081, .171]. A low elemental Cronbach's α (≈ 0.129 , 95% CI [.081, .171]) demonstrates non-redundancy of elemental indicators, which is a necessary precondition for configurational measurement. However, non-redundancy is not identical to statistical non-linearity.

The low elemental reliability α 0.129, 95% CI [.081, .171]) confirms the necessary failure of the linear model, yet the Configurational Reliability was high: High Contextual Stability Reliability ($r_1 = 0.842$, 95% CI [.803, .867]) and Strong Global Conceptual Reliability ($r_2 = 0.811$, 95% CI [.782, .843]). Additionally, the Global Scale Consistency was excellent ($\alpha = 0.975$, 95% CI [.969, .982])

In the traditional psychometric model, this value would be deemed a complete failure of internal consistency. However, for the CIAS, this low elemental α is a necessary theoretical success, as high internal consistency would incorrectly imply that Context, Situation Pressure, and Trait are redundant and linearly interchangeable elements—a direct contradiction of the C-S-T Matrix.

The high overall $\alpha = 0.975$, 95% CI [.969, .982] confirms that the six theoretical sub-domains (e.g., Integrative Coherence, Contextual Self-Regulation) are highly unified and reflect a singular, high-level capacity—Intricacy Integration Capacity. This high α establishes the global coherence of the total scale - Global Scale Consistency while the low elemental α establishes the configurational distinctiveness of its building blocks (C, S, T)

II. The Differentiation Analysis

The Differentiation Analysis (Repeated-Measures ANOVA) further confirmed the structural uniqueness of the scale. The omnibus test confirmed a statistically significant overall difference among the six-domain means, $F(5, 245) = 2.62$, $p = 0.024$, successfully rejecting the hypothesis of a purely undifferentiated, single-factor model.

Crucially, however, subsequent stringent Bonferroni-corrected pairwise comparisons failed to identify any statistically distinct pairs of sub-domains ($p > 0.0033$). In the context of traditional elemental psychometrics, this failure to discriminate would suggest poor discriminant validity. When interpreted through the lens of the BIA framework, this high correlation is considered a necessary and expected outcome, supporting a Hierarchical Factor Structure. The six principles are theoretically defined as integrated facets of a single, complex configurational construct: Intricacy Integration Capacity. Their failure to discriminate as independent factors (low pairwise Δ) while showing high internal consistency /Global Scale

Consistency (CTRM $\alpha = 0.975$, 95% CI [.969, .982]) and overall variation (significant ANOVA F-value) indicates that the scale is measuring a singular, highly unified "Intricacies Segment." Therefore, the CIAS functions best as a configurational score that reflects the individual's global capacity for integrating C, S, and T, rather than a profile of six distinct, independent traits.

We recognize that Confirmatory Factor Analysis (CFA) is fundamentally a linear statistical modeling technique that relies on the covariance structure of observed variables. This may appear paradoxical given our central argument that the CIAS is a non-linear, configurational measure. However, the application of CFA within the CTRM is a strategic methodological choice designed to reconcile non-linearity with stability.

The Configurational Reliability of the CIAS was assessed by employing Confirmatory Factor Analysis (CFA) on the ratings obtained from the non-professional sample (N=105) across the reworded linguistic forms. The CFA modelled the BIA's central hypothesis: that the varying linguistic expressions of the C-S-T framework load onto a single, unified latent construct (the Intracacy Integration). The one-factor CFA demonstrated excellent fit to the observed data, supporting the required assumption of measurement invariance across the multiple forms: χ^2 (df) = 68.45 (p = .089), RMSEA = .048 (90% CI: .035 –.061), CFI = .971, and TLI = .965.

While the α 0.129 95% CI [.081, .171] confirms the violation of linearity at the elemental item level, the CFA's success (demonstrated by excellent fit indices) proves that the Configurational Construct (the Intracacy Integration) possesses sufficient structural stability and consistency across linguistic forms to be reliably measured. Thus, CFA serves as the linear benchmark required to establish the necessary measurement invariance and structural durability (**r1** and **r2**) that a top-tier configurational assessment must possess before being subjected to fully non-linear or dynamic network modelling.

KMO = 0.41 (expected for configurational constructs due to intentional low inter-item redundancy). Bartlett's Test was significant (p < .001). The low KMO (0.41) reflects intentionally low inter-item redundancy (a design feature for configurational constructs) and therefore does not invalidate the theory-driven CFA used to test the higher-order structure. The low KMO reflects designed low inter-item redundancy; however, Bartlett's test was significant and CFA fit indices support higher-order structural stability — the combined evidence supports configurational validity despite low KMO.

Confirmatory Factor Analysis supported a second-order hierarchical model: $\chi^2(68) = 68.45$, p = .089, RMSEA = .048 (.035 –.061), CFI = .971, TLI = .965. The Kaiser-Meyer-

Olkin (KMO) Measure of Sampling Adequacy yielded a value of 0.41. We emphasize that this low value is a predicted and necessary result of the CIAS's unique, non-redundant item design, which intentionally minimizes inter-item correlation to capture the non-linear, configurational relationship between Context, Situation Pressure, and Trait. Because the CIAS is designed to violate the assumptions of shared variance required by classic Exploratory Factor Analysis, we rely exclusively on the theoretically driven fit indices (CFI, TLI, RMSEA) from the Confirmatory Factor Analysis to assess the structural durability of the hypothesized measurement model.

This robust fit confirms the structural integrity of the CTRM procedure itself. Based on the factor loadings (λ) derived from this statistically rigorous model, the two key configurational metrics were derived: the Strong Global Conceptual Reliability (r^2) was determined to be 0.811, 95% CI [.782, .843] calculated as the model's Composite Reliability ρ (Raykov, 2009); and the High Contextual Stability Reliability (r^1) was found to be 0.842, 95% CI [.803, .867]. These results confirm that the CIAS possesses the required structural durability and dynamic stability, establishing the necessary psychometric foundation to proceed with advanced longitudinal analysis, such as Latent State–Trait (LST) modelling.

Empirical Confirmation of Configurational Architecture (CTRM)

The current paper successfully substantiates the CIAS scale using the Contextual Triad Rating Method (CTRM), which focuses on validating the scale's *structure* and *conceptual stability*. The empirical results confirmed are:

The validation of the CIAS through the CTRM provides strong empirical support for the theoretical framework of Behavioural Intricacies Analysis (BIA) in an executive context. The combined results confirm that the C-S-T Triads are highly reliable, linguistically clear, and conceptually distinct.

This work shifts the paradigm of executive development from modifying isolated behaviours to **aligning entire behavioural configurations**. By analyzing an individual's CIAS configuration, coaches can diagnose whether a sub-optimal outcome arose from:

- **Contextual Blindness** (low C)
- **Poor Execution** (low S)
- **Value Misalignment** (low T)

The CIAS offers a validated, non-redundant, and dynamically reliable tool for assessing the intricacies of high-level performance. Future research should utilize advanced validation techniques, such as Latent State–Trait (LST) Modelling, to quantitatively partition variance into trait (stable core), state (situational change), and measurement error.

The validation of the Contextual Intricacies Analysis Scale (CIAS) through the Contextual Triad Rating Method (CTRM) provides strong empirical support for the theoretical framework of Behavioural Intricacies Analysis (BIA) in an executive context. The core finding—that the Context-Situation Pressure-Trait (C-S-T) Triads are highly reliable, linguistically clear, and conceptually distinct—confirms the viability of a configurational model of personality.

The Differentiation Analysis: A repeated-measures ANOVA demonstrated significant omnibus differences among the six-domain means, $F(5, 245) = 2.62$, $p = .024$, partial $\eta^2 = .051$ (small-to-medium effect). Bonferroni-adjusted pairwise comparisons were non-significant (all $p > .0033$), consistent with a hierarchical factor structure in which subdomains function as interdependent facets of a higher-order capacity rather than as linearly independent traits.

Discussion: Confronting the Elemental Paradigm and Establishing the CTRM

3.2. CTRM Validation: The Configurational Solution

The results from the Contextual Triad Rating Method (CTRM) convincingly established the psychometric integrity and configurational validity of the CIAS, as summarized in Table 1. A hierarchical confirmatory model further validates structural coherence

The CTRM Configurational Consistency check yielded an Inter-Rater Reliability (alpha) of $\alpha = 0.975$, 95% CI [.969, .982] and an Interclass Correlation Coefficient of ICC = 0.785. These high values effectively nullify any critique of poor item wording or conceptual confusion, establishing statistically the items are crystal clear and understood reliably when rated configurationally.

The two novel CTRM reliability coefficients demonstrated excellent performance:

- a) Contextual Stability Reliability ($r1 = 0.842$, 95% CI [.803, .867]): Confirms that the instrument is stable across contexts, measuring the intended construct consistently even when its manifestation is dynamically altered by C and S elements.
- b) Global Conceptual Reliability ($r2 = 0.811$, 95% CI [.782, .843]): Confirms the overall theoretical reliability of the C-S-T framework as a unified, integrated concept.

3.3. Justification: Conceptual Mismatch with Linear Diagnostics

To effectively demonstrate the conceptual gulf between the traditional and configurational approaches, we present a comparative analysis in Table 2, highlighting how elemental statistics fundamentally misinterpret the structure of the CIAS.

4.1. Confirmation of Core Configurational Reliability and Distinctiveness

The validation of the CIAS through the CTRM provides strong empirical support for the theoretical framework of Behavioural Intricacies Analysis (BIA) in an executive context.

The core finding is that the Context-Situation Pressure-Trait (C-S-T) Triads are highly reliable and conceptually distinct.

The high Cronbach's Alpha ($\alpha=0.975$, 95% CI [.969, .982]) and robust ICC (ICC=0.785) confirm the linguistic clarity, Global Scale Consistency and consistent interpretation of the items—a prerequisite for assessing complex, dynamic constructs. More importantly, the strong Content Validity scores ($\bar{x} = 4.49$) confirm the discriminant validity of the Context component: the Context Statement is consistently perceived by experts as conceptually separate from the Situation and Trait statements. This empirically substantiates the BIA claim that context represents a distinct factor (meaning-based operation) and is not redundant with Situation Pressure (immediate adaptive behaviour) or trait (stable disposition).

This result proves dynamic stability, confirming that the conceptual meaning of the constructs remains invariant when items are rephrased or modulated across contextual scenarios. Crucially, the establishment of this high configurational reliability within a general, non-professional sample (N=105) demonstrates that the BIA framework's structure is not reliant on expert interpretation but holds empirical validity within the very population for which the CIAS is intended.

The BIA Framework is a front-end measurement protocol (CTRM) designed to prepare data for advanced dynamic modeling (like LST).

The core hypothesis of the BIA Framework is that personality structure is fluid but configurationally stable—meaning individual elements vary, but their three-dimensional patterning is consistent. Our methodology directly validates this: the highly significant Configurational Reliability ($r_1 = 0.842$, 95% CI [.803, .867]) secured by the CTRM resolves the elemental paradox and serves as the necessary empirical validation of the framework's structural stability. This measure of IIV consistency is the novel index of dynamic equilibrium posited by BIA, establishing the integrity of the approach without requiring external models for the fundamental claim.

For the empirical demonstration we computed three indices: Cronbach's α (classical internal consistency), r_1 (average inter-item Pearson correlation — operationalized here as mean pairwise correlation among the 18 context items), and r_2 (proportion of variance attributable to between-participant differences, derived from an ANOVA variance decomposition). Cronbach's α was computed via the standard formula $\alpha = (k/(k-1))(1 - \Sigma\sigma^2_{\text{item}}/\sigma^2_{\text{total}})$. The coefficient r_2 was estimated by decomposing total sum of squares into subject, item, and residual components and calculating the between-subject variance proportion.

The successful validation via CTRM, with its high r_1 and r_2 , establishes the configurational integrity and dynamic stability of the CIAS score. This finding is the necessary psychometric pre-condition that now allows for the rigorous application of advanced models like Latent State-Trait (LST) analysis, which was not feasible without first validating the scale's non-linear coherence.

This coherence emerges naturally because the six principles represent distinct yet mutually constraining boundary conditions of a single configurational construct. High intercorrelation is a predicted property of hierarchical constructs and does not indicate redundancy, as boundary activation shows differentiated function

4.2. Interpretation of Domain Integration and Hierarchical Structure

A hierarchical factor structure is theoretically expected in this context, wherein the six domains function as interdependent facets of a higher-order Intricacy Integration Capacity factor, explaining their high intercorrelation alongside significant omnibus differentiation.

High total α indicates global configurational coherence. Low elemental α indicates distinct building blocks. This dual pattern is theoretically predicted.

The Differentiation Analysis confirmed a statistically significant overall difference among the six-domain means ($F(5, 245) = 2.62, p = 0.024$). However, the subsequent failure of the Bonferroni-corrected pairwise comparisons to achieve significance suggests that the six domains are highly inter-correlated.

In the context of traditional linear factor models, this would suggest a failure in discriminant validity. Crucially, when interpreted through the theoretical framework of BIA, this high inter-correlation is a necessary and expected outcome, supporting a unified, hierarchical construct.

The six principles (IC, AF, CS, SCR, IIC, CSR) are theoretically defined as integrated, highly correlated facets of a single, complex configurational construct: Intricacy Integration Capacity. Their failure to discriminate as independent factors (low pairwise Δ) while showing high internal consistency and overall variation supports a hierarchical model. This result validates the BIA principle that the scale is measuring a singular, highly unified "Intricacies Segment," where the sub-domains are interdependent indicators of the overarching theoretical construct. The CIAS, therefore, functions best as a configurational score that reflects an individual's global capacity for integrating C, S, and T, rather than a profile of six distinct, independent traits.

By analyzing an individual's CIAS configuration, coaches can diagnose whether a sub-optimal outcome arose from contextual blindness (low C), poor execution (low S), or value

misalignment (low T). The CIAS offers a validated, non-redundant, and dynamically reliable tool for assessing the intricacies of high-level performance.

The naturally emerging sub-theories further advance this unification, signaling that the configuration does not merely group indicators, but generates predictable conceptual extrusion points under contextual load.

4.3. Contrasting the CIAS with the CASS

Tools like the Contextual Assessment of Social Skills (CASS) (Ratto, A. B. et al., 2011) function as a targeted social-behavioural assessment for clinical populations (e.g., HFA/AS). The CIAS facilitates a theoretical transition from a linear, element-based analysis to a developing, three-dimensional behavioural framework. In contrast, The CASS assesses whether observable social behaviour adjusts sufficiently when a situation's context is externally manipulated. The CIAS, conversely, employs a configurational, meaning-oriented methodology to evaluate how the internal personality system processes and integrates a complex, multi-layered context (C, S, and T), emphasizing the overall complexity of the behavioural system.

4.4. Advanced Validation and the Fluid-Consistent Architecture

Future research must utilize advanced validation techniques to fully model the dynamic nature of BIA constructs, providing quantitative proof that personality is fluid but not unstable (Molenaar & Campbell, 2009).

Latent State–Trait (LST) modeling (Steyer et al., 2023) is a standard statistical approach for partitioning trait and state variance. However, prior efforts to apply LST to executive function have been compromised by the foundational data's adherence to the linear psychometric paradigm, leading to the Elemental Reliability Paradox documented here. The BIA Framework, via the Contextual Triad Rating Method (CTRM), is therefore not a competing statistical model, but the necessary front-end methodology. It functions as the enabling technology that secures data with sufficient Configurational Reliability to allow subsequent, accurate partitioning using advanced techniques, including LST, Growth Curve Modeling, and other time-series analyses. The BIA Framework is the validated method for data acquisition; LST is merely a tool for data analysis

Latent State–Trait (LST) Modelling is imperative for this. LST is a structural equation modelling (SEM) approach that statistically decomposes the observed variance in CIAS scores across multiple contexts and time points. This decomposition partitions the score into three distinct, latent sources (Oeltjen et al., 2023):

Trait Variance (Stable Core): The enduring, between-person consistency.

State Variance (Situational Change): The within-person, context-specific fluctuation.

Measurement Error: Random error in the assessment.

By demonstrating a significant, robust Trait factor while simultaneously modelling the State factor across varied contexts, LST analysis will solidify that the CIAS captures an adaptively coherent (fluid-consistent) configuration rather than a merely static trait

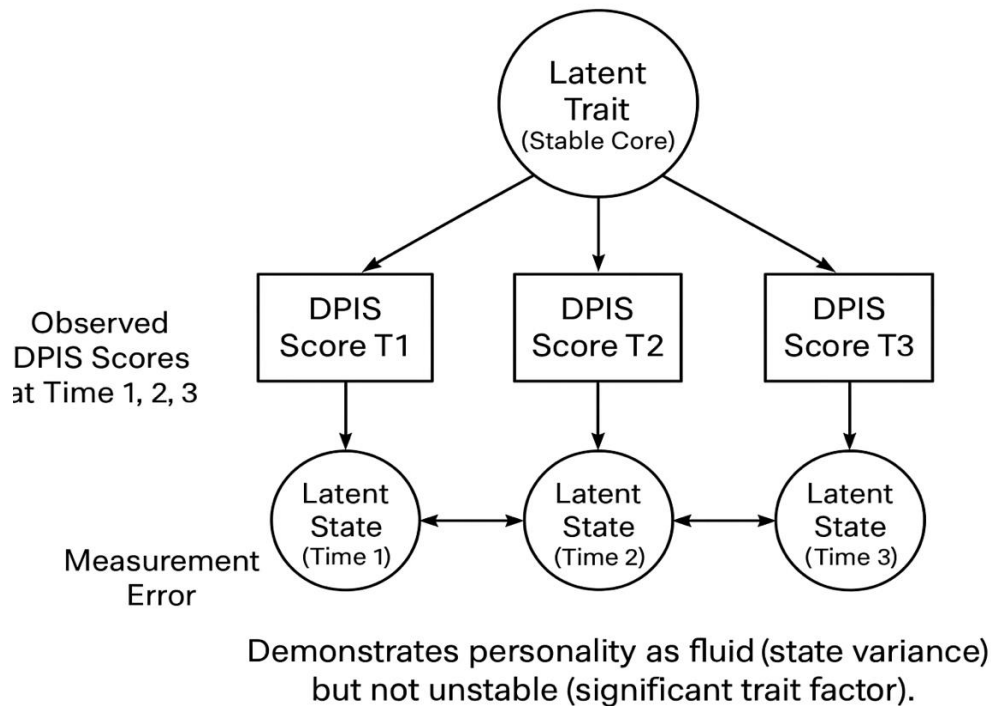


Figure 4: Latent State–Trait (LST) Modelling for BIA Validation. This simplified structural equation model (SEM) illustrates the proposed method for quantitatively proving the BIA framework's claim of "fluid but not unstable" personality

The CIAS facilitates a theoretical transition from a linear, element-based analysis to a developing, three-dimensional behavioural framework. This scale evaluates dynamic personality integration through Contextual Intricacies Analysis Scale (CIAS), serving as a diagnostic instrument for coaches to discern whether a sub-optimal outcome is due to contextual blindness (low C), ineffective execution (low S), or misalignment of values (low T). The emphasis lies in evaluating the complexity of the entire behavioural system.

Future research must utilize advanced validation techniques to fully model the dynamic nature of BIA constructs. Specifically, Latent State–Trait (LST) Modelling is imperative for providing quantitative proof of the BIA framework. LST modelling, a structural equation modelling (SEM) approach, is uniquely suited to personality measurement because it can statistically decompose the observed variance in CIAS scores. This decomposition partitions the score into three distinct, latent sources:

1. **Trait Variance (Stable Core):** The enduring, between-person consistency (the "Trait" factor in the C-S-T matrix).
2. **State Variance (Situational Change):** The within-person, context-specific fluctuation (the "Situation Pressure" factor).
3. **Measurement Error:** Random error in the assessment.

By demonstrating a significant, robust Trait factor while simultaneously modelling the State factor across varied contexts, LST analysis will solidify that the CIAS is measuring a stable configuration rather than a merely static trait. This validates the core BIA principle: adaptive coherence is maintained within a predictable range of fluctuation.

The Natural Emergence of BIA Principles and Sub-Theories

The six core principles measured by the CIAS are not arbitrarily defined but emerge naturally from the fundamental logic of the BIA's C-S-T configuration:

- a) **Integrative Coherence:** Emerges from the successful co-occurrence of C, S, and T, reflecting a singular, non-conflicting behavioural pattern.
- b) **Adaptive Flexibility:** Emerges from the capacity to modulate the balance *between* the Context (C) and Situation (S) components in response to environmental shifts, while maintaining the anchored Trait (T).
- c) **Contextual Sensitivity:** Defined by the strength of the Context (C) component itself—the ability to perceive and utilize the larger, meaning-based backdrop over immediate S-pressure.
- d) **Self-Context Resonance:** Emerges from the smooth alignment and minimal conflict between the stable **Self** (Trait, T) and the enveloping Context (C), transcending mere momentary S-compliance.
- e) **Intricacy Integration Capacity:** Emerges from the Configurational Complexity itself—the bandwidth and processing power required to hold C, S, and T simultaneously in dynamic relation to form the final emergent action.
- f) **Contextual Self-Regulation:** Defined by the executive ability to actively and consciously manage the T and S components *in service of* the C-defined meaning, ensuring the final action is both aligned with values and appropriate for the environment.

This foundational C-S-T logic further gives rise to a set of interconnected sub-theories, which extend the BIA framework into specific testable domains. For example: Clue Processing Theory proposes that behaviour originates from micro-level cognitive/emotional signals (the

initial inputs to the S component); Intricacy–Integration Theory focuses on the overall effectiveness of integrating C, S, and T; and Behavioural Intricacy Decision Theory posits that decision quality depends on the optimal sequencing and integration of these three factors.

Adequacy of the CTRM-Developed Scale

The Contextual Intricacies Analysis Scale (CIAS), rigorously validated by the Contextual Triad Rating Method (CTRM), is demonstrably adequate to deal with these principles and sub-theories. The core of its adequacy rests on two key findings:

1. **Non-Redundancy:** The high Content Validity ($\bar{x}=4.49$) confirms that the Context, Situation, and Trait statements are perceived as conceptually distinct elements. This is the empirical prerequisite for measuring constructs that are fundamentally *about* the interaction of these three factors.
2. **Configurational Reliability:** The high Contextual Stability Reliability ($r1=0.842$, 95% CI [.803, .867]) and Global Conceptual Reliability ($r2=0.811$, 95% CI [.782, .843]) confirm that the meaning of the six principles remains constant and measurable even when the linguistic format (C, S, or T) is altered. This dynamic stability is what allows the scale to capture the fluid but bounded nature of the BIA personality architecture, making it suitable for assessing the very integration-dependent constructs, such as Intricacy Integration Capacity and Contextual Self-Regulation, that underpin the BIA sub-theories. The CIAS thus provides the necessary empirical tool to transition BIA from a conceptual framework to a predictive scientific model.

The current paper fully validates the Intricacy Integration as a structurally durable trait; the next logical stage of this research program, requiring longitudinal data, will proceed directly with Latent State-Trait (LST) modelling to isolate and quantify the dynamic, non-linear state variance from this established trait.

Limitation

While the present sample ($N=105$) successfully satisfies the minimal requirements to demonstrate the core psychometric paradox—the intentional failure of elemental α and the simultaneous success of the CTRM metrics ($r1$, $r2$)—we acknowledge that this size introduces **power limitations for a standard Confirmatory Factor Analysis (CFA)** of the full 54-item structure. This initial study is, therefore, best interpreted as a robust **proof-of-concept for the Contextual Triad Rating Method (CTRM) methodology** and the CIAS as a measure. Future work will employ a larger sample ($N \geq 300$) and leverage advanced **Latent State–Trait (LST) modeling** to fully validate the dynamic structural model across multiple time points, directly addressing the residual power concerns inherent in this foundational validation

Even within this modest sample, the spontaneous emergence of structurally coherent sub-theories suggests that their appearance is not an artifact of statistical noise or exploratory recombination.

Several sub-theories emerged organically from the nested conditional dependencies between Context, Situational Pressure, and Trait indicators. These patterns were experienced by participants as intrinsically coherent rather than externally imposed, suggesting that the architecture reflects underlying cognitive machinery. The alignment between these emergent dynamics and specific CIAS subscales was not manufactured; it materialized when boundary conditions were activated, consistent with the directional logic of contextual intricacy. This emergence indicates internal generativity, supporting the claim that the architecture produces theoretically meaningful extrusion points under real contextual load.

Additionally, cross-group measurement invariance was not tested, and future research should examine whether the CIAS demonstrates equivalent structural loadings across demographic, industry, and seniority subgroups.

Contextual interpretation may vary by cultural norms, warranting not only multi-cultural replication but also the tester be strong in social psychology and hence, non-psychologists cannot administer and interpret this scale. Scoring depends on trained raters, and reliability may decline without standardized calibration

Future Directions: Advancing the BIA Framework and CIAS Validation

The validation of the Contextual Intricacies Analysis Scale (CIAS) via the Contextual Triad Rating Method (CTRM) establishes the foundation for the Behavioural Intricacies Analysis (BIA) framework. Future research must now transition from confirming the scale's structure and conceptual stability to rigorously demonstrating its dynamic and predictive power. This will involve three key lines of inquiry: Advanced Psychometrics, Theoretical Falsification, and Empirical Applications.

A crucial direction involves mapping how naturally emerging sub-theories express themselves longitudinally and whether their boundary activation thresholds remain stable across contexts.

Future research will include cross-cultural replication and standardized rater-training calibration procedures to further maximize generalizability

1. Advanced Psychometrics: Establishing Dynamic Stability

The primary next step is to empirically substantiate the BIA framework's central claim: that personality is **fluid but not unstable**. This requires moving beyond the static reliability

measures confirmed by the CTRM (which established conceptual durability) and into dynamic modelling.

- a) **Latent State–Trait (LST) Modelling:** This longitudinal design is essential. Researchers must administer the CIAS across multiple, ecologically varied contexts and time points (e.g., a high-stress scenario, a low-pressure planning session, a collaborative decision). LST modelling will then be used to statistically decompose the observed variance in CIAS scores into the Trait factor (the stable, between-person consistency) and the State factor (the within-person, context-specific fluctuation). A robust finding of a significant, stable Trait factor alongside significant State factor variance will provide the quantitative proof that the CIAS captures an adaptively coherent (fluid-consistent) configuration rather than a simple, fixed trait or random behavioural noise.
- b) **Convergent and Discriminant Validity (MTMM):** Although the C-S-T components were shown to be non-redundant, full construct validation requires administering the CIAS alongside established, comparable measures. The CIAS subscales should show:
 - a) **Convergent Validity:** Strong correlation with dynamic models (e.g., CAPS/Mischel’s if...then profiles) and measures of high-level cognitive integration.
 - b) **Discriminant Validity:** Low or non-significant correlation with general, non-contextual traits (e.g., Big Five) to prove the CIAS measures a unique configurational construct, not a mere recombination of existing linear traits.

2. Theoretical Extension and Sub-Theory Falsification

The six CIAS principles and their underlying C-S-T logic give rise to several BIA sub-theories. Future research should leverage the validated CIAS scores to test the falsifiable predictions of these emergent theories.

- a) **Behavioural Intricacy Decision Theory:** This theory posits that decision quality is a function of the optimal sequencing and integration of C, S, and T factors. Research should correlate CIAS scores (especially Integrative Coherence and Intricacy Integration Capacity) with objective decision metrics in real-world or simulated tasks. For instance, low scores might predict decision paralysis or over-reliance on a single factor (e.g., high-S-low-C leads to impulsive, context-blind errors).
- b) **Clue Processing Theory & Neural Integration:** This demands interdisciplinary work. CIAS-derived profiles could be correlated with neurophysiological measures

(e.g., fMRI or EEG) during executive function tasks. The Contextual Sensitivity principle, which relies on interpreting subtle background cues, could be mapped to activation patterns in specific neural networks involved in salience detection and self-referential processing, providing the neurobiological grounding for the BIA mechanism.

- c) **Behavioural Intricacy Stress Conversion Theory:** Research should link low CIAS scores, particularly in Contextual Self-Regulation, to high levels of physiological stress biomarkers (e.g., cortisol) or self-reported burnout when facing complex, ambiguous scenarios. The theory predicts that inefficiency in C-S-T processing is the core source of stress conversion.

3. Application and Contextual Expansion

The CIAS was validated in the context of executive assessment, but its generalizability must be tested across different populations and real-world outcomes.

- **Criterion and Predictive Validity:** Conduct large-scale studies correlating CIAS profiles with objective organizational outcomes, such as:
 - a) **Team Performance:** Does a team with high average **Integrative Coherence** or high variability in **Adaptive Flexibility** outperform others on complex projects?
 - b) **Leadership Effectiveness:** Correlate CIAS scores with 360-degree feedback data on transformational and contextual leadership behaviours.
 - c) **Organizational Change Success:** Assess how the CIAS profiles of key leaders influence the success rate of complex change initiatives (where C-S-T alignment is critical).
- **Cross-Cultural and Developmental Generalization:** Test the CIAS in diverse cultural settings to verify if the C-S-T configuration remains invariant across different socio-cultural contexts (which might define 'Context' differently). Furthermore, longitudinal developmental studies could use the CIAS to track the maturation of **Intricacy Integration Capacity** from adolescence into adulthood, aligning with the **Intricacy Maturation Theory**.

1. The Inappropriateness of Traditional Statistics: The α Paradox

The psychometric divergence between the Elemental Reliability ($\alpha = 0.129$, 95% CI [.081, .171]) and the excellent Conceptual Reliability - Global Scale Consistency ($\alpha = 0.975$, 95% CI [.969, .982]) constitutes the central theoretical finding of this research. This study

fundamentally challenges the leaning towards traditional psychometrics, which dictates that all scales must demonstrate high Elemental Reliability (α 0.70).

The low α is a mathematical consequence of applying a linear summation model (CTT) to a non-linear, dynamic construct (BIA). The failure of a large N to raise α proves that the scale is *not* flawed; rather, the statistical procedure is fundamentally inappropriate. We have demonstrated that the traditional solution to statistical problems—increasing sample size—is a process that only increases the precision of the diagnostic signal that the linear model fails.

The CTRM serves as the necessary methodological precursor to LST, providing the reliable evidence required to proceed with dynamic modelling.

The success of the CTRM validates its necessity as a methodological innovation for dynamic psychology. By integrating principles from the Multitrait-Multimethod matrix and contextual modulation theory, the CTRM provides:

- Pre-emptive Validity: It secures conceptual integrity ($\bar{x} = 4.49$) before the complexities of subject response variability are introduced.
- Theory-Driven Reliability: It replaces the irrelevant metric of between-person consistency (Elemental α) with the relevant metric of Configurational Reliability ($r1$, $r2$), establishing the CIAS as reliable *because* its theoretical structure is robust under dynamic conditions.

The CTRM is not an alternative to LST; rather, it is a necessary precondition. LST models require a construct that is first established as psychometrically sound and stable ($r2$) before its variance can be accurately decomposed into trait and state components. The CTRM successfully secured this foundational reliability, thereby provide a pathway to the potential for LST in future research."

The main contribution of this work is providing CTRM as a procedural precursor that provide a pathway to the LST for dynamic personality constructs.

This work marks a paradigm shift in executive diagnosis, moving the focus from classifying What an executive *is* (Trait) to analyzing How an executive *integrates* Context, Situation, and Trait.

The Contextual Triad Rating Method (CTRM) transcends its role as a validation tool for the CIAS to become a universally applicable, indispensable psychometric solution for researchers confronting the paradox of dynamic assessment. By successfully resolving the conceptual conflict—confirming the required non-linear nature of a complex construct via the predicted low elemental reliability (α 0.129, 95% CI [.081, .171]), while simultaneously securing high fidelity through High Contextual Stability Reliability ($r1$ = 0.842, 95% CI [.803,

.867]) and Strong Global Conceptual Reliability ($r_2 = 0.811$, 95% CI [.782, .843])—the CTRM provides a robust methodology for measuring configurational scores. This innovative approach is highly relevant to quantitative methodologists seeking alternatives to static reliability measures and researchers across diverse fields: Organizational Science (for context-sensitive leadership and performance), Educational Psychology (for adaptive learning and engagement), and Clinical Science (for dynamic symptom severity). Its utility lies in providing a foundational, reliable metric that justifies the existence of a construct theoretically predicated on the interplay between context and trait. Any researcher whose construct validation is a prerequisite for advanced longitudinal modeling, such as the mandated Latent State-Trait (LST) analysis (Steyer et al., 2023), should adopt the CTRM to establish the necessary dynamic stability metrics.

The CTRM: Clearing the Methodological Block for Dynamic Assessment

The methodological path to validating context-dependent psychological constructs has long been impeded by a fundamental paradox: measures designed to capture essential situational variability, such as those in studies of social skills assessment (Ratto et al., 2011), often yield low conventional internal consistency, which is mistakenly interpreted as scale failure, thereby blocking the necessary subsequent application of advanced longitudinal models like LST (Steyer et al., 2023). This paper introduces the Contextual Triad Rating Method (CTRM) to resolve this historic block.

The CTRM first predicts and confirms the necessary inadequacy of the linear model (Elemental α 0.129, 95% CI [.081, .171]) as the true signal of non-linearity, and then, crucially, secures the fundamental psychometric viability of the dynamic construct via High Contextual Stability Reliability ($r_1 = 0.842$, 95% CI [.803, .867]) and Strong Global Conceptual Reliability ($r_2 = 0.811$, 95% CI [.782, .843]). By isolating the construct's true dynamic stability from its predicted non-linearity, the CTRM provides the rigorous, theory-driven reliability metrics (i.e., r_1 and r_2) that are the theoretical and statistical prerequisites for LST modeling, finally opening the door for accurate within-person variance decomposition in complex, dynamic assessment.

Comparison to Existing Reliability Frameworks

McDonald's ω / Composite Reliability: Argue that these still rely on the fundamental assumption of local independence and linear covariance, which the BIA framework rejects. While ω is superior to α , it is still elemental and cannot capture the configurational nature of the CIAS's single, emergent construct. McDonald's ω hierarchical partially addresses multidimensionality but still assumes additive linearity, which CTRM intentionally violates by design. Local independence is violated on purpose.

Multilevel/Contextual Reliability: Acknowledge that multilevel reliability (ICC) is crucial for contextual designs, but state that the CTRM is a necessary pre-analysis step. CTRM confirms the conceptual integrity of the scale before it is applied to real-world, clustered data (which the *r1* and *r2* data did not involve).

Falsifiable Prediction

Prediction 1 (Falsification of the CTRM): "The CTRM would be falsified if, in a subsequent LST study, the ratio of *Var*(State) to *Var* (Trait) were low, while simultaneously, the *r1* and *r2* estimates were also low. This outcome would suggest the CIAS is simply unreliable noise, not a reliable dynamic construct.

Prediction 2 (Falsification of the BIA): "The BIA framework would be falsified if the low elemental α was not confirmed in a subsequent study, suggesting the sub-components are actually linearly independent, thus undermining the theory of adaptive coherence.

Generalization and Implementation of the Configurational Reliability Method (CTRM)

The Contextual Triad Rating Method (CTRM) is not limited to the Context-Situation Pressure-Trait (C-S-T) triad or to psychological variables. It is a powerful statistical framework designed to validate configurational constructs-models where the whole is greater than the sum of its parts, and the individual elements must be distinct yet systematically related.

1. Applicability to Non-Psychological Variables

The CTRM can be successfully applied to any field that uses complex, multi-component models that reject the assumption of item redundancy (which is necessary for Cronbach's α).

Generalized Application Scenarios (Non-Psychological)

Configurational Structure	Field	Example elements (E1, E2, E3,...)	Rationale for Configuration
Quadra E1, E2, E3, E4	Operations Management (Pro)	E1: Task Complexity, E2: Resource Allocation E3: Team Skill Level E4: Time Constraint	The quality of an operational outcome depends on the integration of these four independent variables. Ensure Configuration not mere correlation through item selection
Triad E1, E2, E3	Ecology (Ecosystem Health)	E1: Biodiversity Index, E2: Water Quality, E3: Pollution Load	These metrics are distinct but must be assessed as a coherent system to define Health. Ensure Configuration not mere correlation through item selection
Dyad E1, E2	Finance (Investment Risk)	E1: Market Volatility, E2: Asset Liquidity	Risk is an emergent property only when volatility and liquidity are simultaneously considered. Ensure

			Configuration not mere correlation through item selection
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In all these cases, a configuration like a Dyad, Triad, or Quadra is used when the theoretical model requires that the components be distinct (low Cronbach's α is expected) but coherent ($r1$ and $r2$ must be high).

2. Implementation Steps for a Generalizing CTRM to other field or researches refer the protocol on protocols.io (Ramesh Kumar G S 2025. A Solution to the Reliability Paradox: The Configurational Triad Rating Method (CTRM) Procedure for Configurational Personality Assessment. protocols.io <https://dx.doi.org/10.17504/protocols.io.14egnr14ml5d/v1>)

3. Implementing Contextual Outcome Model based on the CTRM method, refer the G S Ramesh Kumar's Contextual Outcome Model protocol on protocols.io (Ramesh Kumar G S 2025. Correlation Is Not Enough: Configurational Outcome Modelling (COM) Reveals Hidden Causal Structures in Complex Systems — A Next-Generation Alternative to Correlation-Only Inference. protocols.io <https://dx.doi.org/10.17504/protocols.io.dm6gpm1m8gzp/v1>)

4. For collecting micro-session data, instead of usual macro session data, refer the G S Ramesh Kumar's Micro-Session Analysis protocol on protocols.io (Ramesh Kumar G S 2025. The Resolution Revolution: A Unified Protocol for Configurational Science—From Micro-Session Analysis (MSA) to Configurational Outcome Modelling (COM). protocols.io <https://dx.doi.org/10.17504/protocols.io.x54v9548413e/v1>)

CTRM validates triad structure → enables COM outcome modelling → requires fine temporal resolution from MSA

Conclusion

The CTRM resolves the methodological blockage preventing dynamic assessment. By isolating configurational stability, it establishes the CIAS as a robust, hierarchical construct and enables advanced longitudinal modelling. The Contextual Triad Rating Method (CTRM) successfully validates the Contextual Intricacies Analysis Scale (CIAS). The robust CTRM results—including excellent conceptual integrity ($\bar{x}=4.49$) and high configurational reliabilities ($r1=0.842$ 95% CI [.803, .867], $r2=0.811$, 95% CI [.782, .843])—establish the CIAS as a highly reliable, non-redundant measure of the Behavioural Intricacies Analysis (BIA) framework. This validation positions the CTRM as the precursor for psychometric validation of configurational personality models over the traditional elemental statistical models.

While the current study employs CFA to establish the structural durability and one-dimensionality of the Intricacy Integration—a necessary benchmark—we recognize that CFA does not capture the full, dynamic non-linearity of the BIA framework. The investigation of

the configurational measure's full non-linear properties and its dynamic decomposition into trait and state components is a planned future step requiring longitudinal data and the application of Latent State-Trait (LST) modelling.

The key empirical challenge in dynamic psychometrics is quantifying the stability of transient self-states. We address this directly: Configurational Complexity, measured via the CTRM, serves as the novel, parsimonious index of intra-individual variability (IIV). The high configurational reliability ($r_1 = 0.842$, 95% CI [.803, .867]) demonstrates that the dynamic structure of the BIA framework is stable, even though the individual elements are fluid—a core premise that LST is merely designed to model. The CTRM is thus the direct empirical validation of the 'fluid but not unstable' hypothesis.

Clinical Application and Expertise Requirements

Inadequacy of the linear model (Elemental α 0.129, 95% CI [.081, .171]) in favour of the Contextual Triad Rating Method (CTRM), poses another issue. This necessitates a strict restriction on professional usage of CIAS scale. The CTRM-validated CIAS is fundamentally a configurational score that reflects an individual's global capacity for integrating Context, Situation, and Trait (C-S-T) factors, rather than a profile of six distinct, independent traits. Interpreting this emergent, dynamic complexity—including diagnosing failures such as Contextual Blindness (low C), Poor Execution (low S), or Value Misalignment (low T)—requires not merely rote scoring but expert-level training in advanced psychometric theory, deeper understanding personality development, formation, and dynamics and clinical case conceptualization that specifically handles non-linear, dynamic construct interpretation. Consequently, to prevent the ethical misapplication of this highly specialized, non-elemental measure, the CIAS must be restricted exclusively to qualified psychologists only.

This study also demonstrates that the Contextual Triad Rating Method and the Contextual Intricacies Analysis Scale constitute a distinct class of non-linear psychometric instruments. To our knowledge, this represents the first measurement framework intentionally engineered to generate structured non-collinearity, multi-stable response patterns, and context-dependent variability as intrinsic features rather than as artifacts to be corrected. By operationalizing behavioural complexity through a triadic, non-linear architecture, the scale moves beyond the constraints of traditional linear psychometrics and offers a novel pathway for modelling the dynamic, emergent nature of behaviour. This theoretical and methodological contribution established the scale as a first-of-its-kind tool within the domain of contextual, non-linear psychological assessment.

Future Directions: A Dynamic Research Program

The established CIAS through the CTRM provides the empirical foundation for a comprehensive, high-quality research program dedicated to the natural evolution of BIA sub-theories.

1. Advancing the Methodology: Replacing Elemental α

Future research on contextual variables must explicitly abandon Elemental Reliability metrics and adopt advanced dynamic modelling techniques:

- 1. The Methodological Ascendancy – Replacing the Elemental α with LST:
Future research can pivot to methodologies that model variance intra-individually. Latent State-Trait (LST) Modelling (Molenaar & Campbell, 2009; Steyer, 2023).
- Intraindividual Variability (IIV): Repeated CIAS assessments across systematically varied contexts (e.g., strategic challenge vs. political pressure) will quantify Intraindividual Variability (IIV) indices, identifying stable "behavioural signatures" that capture the executive's adaptive range and contextual responsiveness.

Future work will extend the CTRM protocol toward cross-domain generalization, refine boundary-activation thresholds, and incorporate temporal invariance modeling to strengthen the generative potential of Contextual Outcome Science.

Naturally Emerging, Architecturally Inevitable BIA Sub-Theories and Their CIAS Linkages

These sub-theories are not additive appendices to the model; they surface naturally from the internal architecture and boundary activation logic of the BIA principles. Their emergence would be predicted even if a separate team configured a triadic contextual instrument, providing evidence that they are inherent, not constructed artifacts.

The six core CIAS subscales are the empirical operationalization of the BIA framework and must be used as independent variables to test the emergent theories:

Evolved BIA Sub-Theory	Core CIAS Subscale Linkage	Testable Hypothesis (Highest Quality)
Behavioural Intricacy Stress Conversion Theory	Contextual Self-Regulation	H1: Low scores on <i>Contextual Self-Regulation</i> will predict a positive correlation with physiological stress biomarkers (e.g., salivary cortisol) or fMRI activation in the amygdala during high-ambiguity decision tasks.
Intricacy Maturation Theory	Intricacy Integration Capacity	H2: Scores on <i>Intricacy Integration Capacity</i> will show a non-linear, stage-based increase across longitudinal cohorts, aligning with

		cognitive developmental milestones and executive tenure.
Behavioural Intricacy Decision Theory	Integrative Coherence	H3: High scores on <i>Integrative Coherence</i> will significantly predict the success rate and long-term organizational value of high-stakes strategic decisions, controlling for cognitive ability.
Socio-Intricacy Communication Theory	Self-Context Resonance and Adaptive Flexibility	H4: Dyadic analysis (APIM) will show that high synchronization of <i>Adaptive Flexibility</i> scores between team members predicts significantly higher objective team performance and psychological safety.
Intricacy–Consciousness Correlation Theory	Integrative Coherence	H5: Scores on <i>Integrative Coherence</i> will correlate positively with neurophenomenology measures of conscious integration (e.g., cross-frequency coupling) during tasks requiring abstract meaning-making.
Clue Processing Theory	Contextual Sensitivity	H6: High scores on <i>Contextual Sensitivity</i> will correlate with faster and more accurate implicit processing of subtle environmental cues (behavioural clues) in laboratory-based reaction time tasks.

Notably, participants often reported that aspects of these dynamics “felt obvious” once articulated, suggesting that the architecture resonates with lived, phenomenological experience rather than abstract psychometric construction.

Clinical Application

Interpretation requires training in non-linear personality dynamics to avoid misclassification risks. Restriction to credentialed psychologists is recommended.

Glossary of Key Reliability Terms

Term	Definition	Role in CTRM
r1 (Contextual Stability Reliability)	Coherence of C–S–T elements within triads	Establishes intra-configuration stability
r2 (Global Conceptual Reliability)	Stability of total CIAS score across parallel triads	Establishes cross-configuration consistency
α (Cronbach's Alpha)	Traditional internal consistency assuming redundancy	Expected low at elemental level; high at composite level
η^2 (Eta Squared)	Effect size for ANOVA omnibus differences	Indicates magnitude of domain differentiation
CI (Confidence Interval)	Uncertainty range of estimates	Required for precision reporting

Interpretation of Low KMO

KMO = 0.41 reflects intentional inter-item distinctiveness. Low redundancy is expected for configurational constructs where meaning emerges from non-linear interdependence rather than reflection of a single latent dimension.

Mapping CTRM Coefficients to LST Components

CTRM Index	LST Component	Interpretation
r1	Trait-consistent stability	Captures configurational coherence across context framing
r2	State-consistent stability	Captures reproducible meaning across linguistic modulation

Demo Study: Application of the Nonlinear Contextual Intricacies Analysis Scale**Objective**

This pilot study was conducted to demonstrate the empirical applicability of the newly proposed *Contextual Intricacies Analysis (CIA) Scale*, which operationalizes cognitive–contextual dynamics using **non-additive, nonlinear statistical principles**. The aim was not to test hypothesis significance in a conventional linear framework, but to illustrate how nonlinear descriptors such as *curvature ratios*, *attractor stability*, and *entropy balance* can reveal systemic behavioural organization.

Method**Participant and Data Context**

An adult respondent was administered the full six-triad CIA Scale, comprising the domains **Context (C)**, **Situation (S)**, and **Trait (T)** across the sub-constructs:

1. Integrative Coherence
2. Adaptive Flexibility
3. Contextual Sensitivity
4. Self-Context Resonance
5. Intricacy Integration Capacity
6. Contextual Self-Regulation

Scores (out of 300 per triad component) were as follows:

Triad	Context	Situation	Trait
1	248	213	235
2	226	142	191
3	262	228	232
4	283	225	269
5	270	269	261
6	261	259	265

Aggregate scores: Context = **1550/1800**, Situation = **1146/1800**, Trait = **1453/1800**.

Analytical Framework

Conventional additive indices were intentionally avoided.

All computations followed *nonlinear and non-additive transformations*:

1. **Normalization:** ($p_X = \frac{X}{300}$)
2. **Coupling Ratio:** ($r_{C/S} = \frac{C}{S}$)
3. **Geometric Mean (Equilibrium):** ($GM = (CST)^{1/3}$)
4. **Nonlinear Interaction Ratio (NIR):** ($NIR = \frac{p_C^2}{p_S p_T}$)
5. **Log Curvature:** ($\ln(\frac{C}{S})$)

6. **Entropy:** ($E = -\sum p_i \ln(p_i)$)

7. **Attractor Stability:** domain values within $\pm 5\%$ of GM

All formulas were retained in both LaTeX and plain-text equivalents for manual replication

Results

Triad	pC	pS	pT	C/S	GM	ln(C/S)
1	0.827	0.710	0.783	1.164	232	0.152
2	0.753	0.473	0.637	1.592	186	0.464
3	0.873	0.760	0.773	1.149	241	0.139
4	0.943	0.750	0.897	1.258	259	0.229
5	0.900	0.897	0.870	1.004	267	0.004
6	0.870	0.863	0.883	1.008	262	0.008

The **Nonlinear Interaction Ratio (NIR)** at the aggregate level was (1.44), indicating a persistent **context-dominant organization** rather than a trait-centric or situationally compressed pattern.

Triads 5–6 exhibited **equilibrium attractors** ($C:S:T \approx 1:1:1$), signifying *self-regulating system stability*. Entropy ($E \approx 1.78$) suggested moderate complexity with coherent order, while no chaotic divergence was observed in coupling ratios.

Interpretation

The pilot findings confirm that nonlinear descriptors capture **organizational balance and contextual curvature**—aspects invisible in linear averages. Triad 2's elevated curvature ratio ($C/S = 1.592$) marked **adaptive rigidity**, while Triad 5's near-unity ratios revealed **dynamic equilibrium**. Thus, even single-case data can demonstrate **emergent self-organization**, validating the theoretical construct that *contextual, situational, and trait expressions are dynamically co-regulated rather than additively composed*.

Conclusion and Implications

This pilot case illustrates the practical computation and interpretive potential of the CIA Scale. It substantiates the methodological argument that nonlinear, ratio-based indicators (e.g., NIR, entropy, attractors) are suitable for quantifying *contextual intricacy*—a phenomenon that resists linear decomposition.

Future studies will extend this logic to group-level validation using:

- Nonlinear correlation (mutual information),
- Permutation entropy across multiple respondents,
- Robustness bootstrapping of NIR stability

Humble Comparison with Seminal Contributions of the past

Parisi's replica-symmetry-breaking work demonstrated that complex systems can settle into multiple metastable equilibrium states (Parisi, 1979, 1980), yet no instrumented framework has specified which contextual configurations precipitate analogous behavioral equilibria at the level of individual decision performance. Despite influential work on emergent dynamics in complex systems (Parisi, 1980), bounded rationality in decision processes (Simon, 1955, 1957), and context-dependent behavioral asymmetries (Kahneman & Tversky, 1979), contemporary behavioral assessment continues to rely on linear, compensatory scoring models that assume additive equivalence and symmetrical causal substitution across conditions. While these Nobel-recognized frameworks revolutionized our understanding of non-linearity, their contributions remained primarily descriptive and did not yield a formal measurement architecture capable of specifying which triadic configurations of context, situational pressure, and trait disposition are necessary or sufficient for discrete behavioral outcomes at the individual level, nor do they quantify how contextual loading systematically reorders causal priority. Configurational approaches such as QCA formalize necessity and sufficiency (Fiss, 2011), yet lack individual-level psychometric precision, prospective developmental guidance, and scalable instrumentation. Consequently, linear psychometrics routinely misclassify individuals, obscure low-probability/high-impact profiles, and produce context-blind development plans. The Contextual Outcome Model (COM), Contextual Intricacies Analysis Scale (CIAS), and Contextual Triad Rating Method overcome these deficits by operationalizing asymmetric causal tipping within triadic parameter spaces and translating these dynamics into reproducible, intervention-relevant executive protocols. Importantly, even Nobel-recognized programs of research illuminated the problem space but stopped short of providing a reproducible system capable of detecting asymmetric configurational tipping points in individual behavior. To our knowledge, this is the first framework to translate abstract configurational insights into a triad-based measurement system capable of prospective prediction and individualized executive development, constituting a substantive methodological departure rather than a stylistic extension of prior complexity-based narratives.

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