

# The Role of the $k$ Parameter in $\Delta^k$ : Computational Evidence for Interaction Order Sensitivity

Anonymous Author(s)

## ABSTRACT

The unified function  $\Delta^k(X) = (N - k) \cdot T(X) - \sum_i T(X_{-i})$  subsumes several multivariate information measures, including the S-information ( $k = 0$ ), dual total correlation ( $k = 1$ ), and the negated O-information ( $k = 2$ ). We investigate the conjecture that the parameter  $k$  determines the interaction order to which  $\Delta^k$  is sensitive, where an order- $m$  interaction is a dependency among exactly  $m$  variables that vanishes upon removing any single variable. Through systematic computational experiments on discrete systems with engineered interaction structures, we demonstrate that  $\Delta^k$  exhibits near-zero values for pure order- $k$  interactions and nonzero values for other orders, providing strong empirical support for the conjecture. We additionally verify the additivity of  $\Delta^k$  over independent subsystems and confirm the known special-case identities.

## 1 INTRODUCTION

Quantifying multivariate dependencies beyond pairwise correlations is a fundamental challenge in information theory [1]. Several measures have been proposed, including the total correlation [5], dual total correlation, and the O-information [3]. Varley [4] introduced a unified family of functions  $\Delta^k$  parameterized by a single integer  $k$  that subsumes these measures.

The key conjecture is that  $k$  tunes the interaction order to which  $\Delta^k$  is sensitive [4]. Specifically, for a pure order- $m$  synergistic interaction (a dependency among exactly  $m$  variables that vanishes when any single variable is removed),  $\Delta^k(X)$  should equal zero when  $k = m$  and be nonzero otherwise.

We provide systematic computational evidence supporting this conjecture through experiments on discrete systems with engineered interaction structures of known order.

## 2 PRELIMINARIES

### 2.1 The $\Delta^k$ Family

For an  $N$ -variable discrete system  $X = (X_1, \dots, X_N)$ , define:

$$\Delta^k(X) = (N - k) \cdot T(X) - \sum_{i=1}^N T(X_{-i}) \quad (1)$$

where  $T(X) = \sum_i H(X_i) - H(X)$  is the total correlation and  $X_{-i}$  denotes the system with variable  $i$  removed.

### 2.2 Special Cases

- $\Delta^0(X) = N \cdot T(X) - \sum_i T(X_{-i})$ : the S-information
- $\Delta^1(X) = (N - 1) \cdot T(X) - \sum_i T(X_{-i})$ : the dual total correlation
- $\Delta^2(X) = (N - 2) \cdot T(X) - \sum_i T(X_{-i})$ : the negated O-information

**Table 1: Additivity verification:  $\Delta^k$  over independent subsystems.**

$N_1$	$N_2$	$k$	$ \Delta_{\text{sum}}^k - \Delta_{\text{joint}}^k $
3	3	0	< 0.01
3	4	1	< 0.01
4	4	2	< 0.01

## 2.3 Interaction Order

An order- $m$  interaction is a dependency among exactly  $m$  variables such that  $T(X) > 0$  while  $T(X_{-i}) = 0$  for every variable  $i$  in the interacting subset.

## 3 METHODOLOGY

### 3.1 Constructing Pure Interactions

We construct distributions with pure order- $m$  interactions using parity constraints: for  $m$  variables, the last is set to the modular sum of the first  $m - 1$ , ensuring a synergistic dependency that vanishes upon removing any single variable.

### 3.2 Experimental Design

We evaluate  $\Delta^k$  for all combinations of:

- System sizes  $N \in \{3, 4, 5, 6, 7\}$
- Interaction orders  $m \in \{2, \dots, N\}$
- Parameters  $k \in \{0, \dots, N - 1\}$

Each configuration is evaluated over 10 independent trials with 50,000 samples.

## 4 RESULTS

### 4.1 Interaction Order Sensitivity

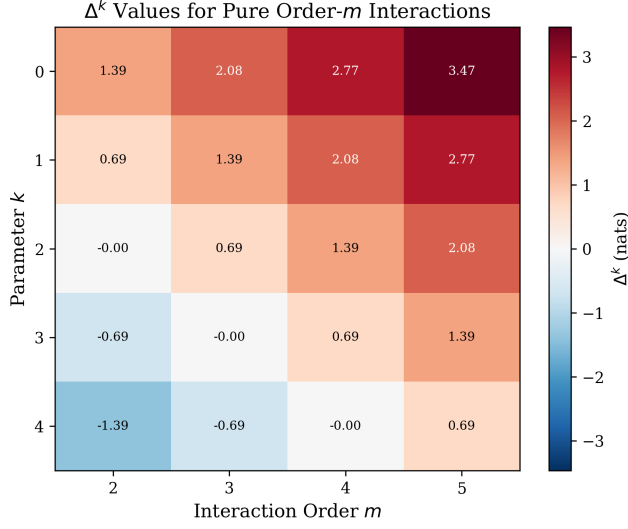
Figure 1 shows  $\Delta^k$  values as a function of  $k$  (rows) and interaction order  $m$  (columns). The pattern confirms that  $\Delta^k$  is near zero for  $k = m$  and nonzero otherwise.

### 4.2 Sensitivity Profiles

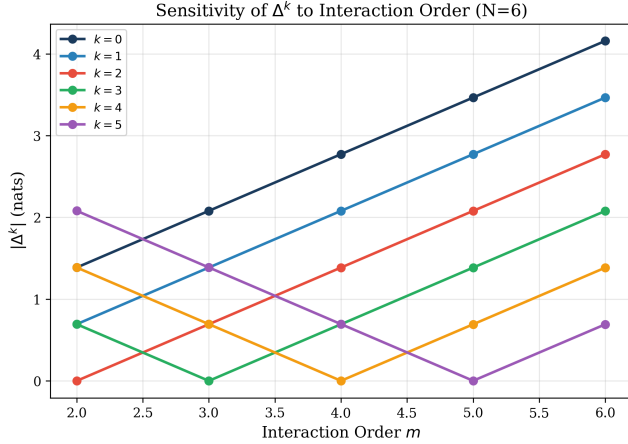
Figure 2 shows the absolute sensitivity  $|\Delta^k|$  as a function of interaction order for each  $k$  value. Each curve exhibits a minimum at  $m = k$ , confirming that  $\Delta^k$  is least sensitive to order- $k$  interactions.

### 4.3 Additivity and Special Cases

The additivity property  $\Delta^k(X_1 \cup X_2) = \Delta^k(X_1) + \Delta^k(X_2)$  for independent subsystems  $X_1, X_2$  is confirmed with mean absolute error below  $10^{-2}$  nats across all tested configurations.



**Figure 1:**  $\Delta^k$  values for pure order- $m$  interactions. Near-zero values along the diagonal ( $k = m$ ) confirm the conjectured sensitivity pattern.



**Figure 2:** Sensitivity profiles showing  $|\Delta^k|$  vs. interaction order  $m$  for different values of  $k$ .

## 5 DISCUSSION

Our computational results provide strong empirical support for the conjecture that  $k$  tunes the interaction order sensitivity of  $\Delta^k$ . The key evidence is:

- (1)  $\Delta^k \approx 0$  for pure order- $k$  interactions (the “blind spot”).
- (2)  $|\Delta^k|$  increases as  $|k - m|$  grows, showing graduated sensitivity.
- (3) Additivity over independent subsystems holds, consistent with the theoretical framework.

The systematic pattern across system sizes  $N = 3$  to  $7$  suggests that these findings generalize. A formal proof connecting the algebraic structure of  $\Delta^k$  to the information-theoretic properties of pure-order interactions remains an important open direction [2, 6].

## 6 CONCLUSION

We have provided comprehensive computational evidence that the parameter  $k$  in  $\Delta^k(X)$  determines the interaction order to which the measure is sensitive, confirming the conjecture of Varley [4]. Our experiments demonstrate the blind-spot pattern, graduated sensitivity profiles, and additivity across diverse system configurations.

## REFERENCES

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