



# Human vs. Computer: Scoring a Test of Emotional Awareness

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## ABSTRACT

Emotional intelligence is the ability to perceive, understand, use, and manage emotions (Mayer et al., 2003). One key aspect of emotional intelligence is emotional awareness: the ability to describe feelings in oneself and others (Lane & Schwartz, 1987). The Levels of Emotional Awareness Scale (LEAS; Lane et al., 1990) includes 20 open-ended items. For each, participants read a short scenario that involves themselves and another person. Then they describe how they and the other person would feel. These written responses are scored based upon the type and number of emotion words used.

Traditionally, the LEAS is scored by hand. However, hand scoring is time-consuming. Therefore, Program for Open-Ended Scoring (POES) was created to score the LEAS (Leaf, 2003). The purpose of this study was to compare the validity of hand scoring and the eight POES scoring methods, by correlating these scores with the 141-item Mayer-Salovey-Caruso Emotional Intelligence Test (MSCEIT; Mayer et al., 2003). A total of 356 undergraduates completed the LEAS and MSCEIT online.

All correlations were significant, thus supporting the validity of the LEAS. HighestN and HighestNUnique had the highest correlations, making these methods a viable alternative to hand scoring. The new AllSumUnique method had a higher correlation than the AllSum method, and the new HighestNUnique method had a higher correlation than the HighestN method, thus indicating that duplicate words add little to our knowledge of someone's emotional awareness. The new CountCat method counts the number of distinct categories used. It deserves further exploration, perhaps with revised categories that more precisely examine the richness of one's emotional descriptions.

## INTRODUCTION

Emotional intelligence is the ability to perceive, understand, and manage emotions and to use emotions to facilitate thinking (Mayer, Salovey, & Caruso, 2004). One key aspect of emotional intelligence is emotional awareness. Emotional awareness is a cognitive ability that allows individuals to identify and describe feelings in themselves and others (Lane & Schwartz, 1987). Emotional awareness is important because recognizing emotion and understanding emotion is something that we do in our day to day functioning.

The Levels of Emotional Awareness Scale (LEAS; Lane, Quinlan, Schwartz, Walker, & Zeitlin, 1990) is an open-ended measure of emotional awareness. The LEAS contains 20 emotionally evocative scenarios that involve the self and another person. For each scenario, participants describe how they would feel and how the other person would feel. Participants get higher scores when they use a larger number of emotion words and more specific emotion words.

The LEAS can be scored by hand or by computer. Scoring a LEAS response by hand is time-consuming because it requires expert subjective judgments based upon the meaning of words, the context in which they appear, and the presence of synonyms. Scoring 100 participants (a reasonable sample size for most psychological studies) would take at least 17 hours.

To score the LEAS with the computer, one can use Program Open-Ended Scoring (POES). The initial scoring methods had strong correlations with hand scoring, high internal consistency, and validity that was comparable to or higher than the validity of hand scoring (Barchard et al., 2010). Recently, a new version of POES has been developed (Ermini Leaf & Barchard, 2013). The AllSum and HighestN techniques remain the same as in the previous version of POES. However, the remaining scoring methods have been revised to be more flexible. First, the AllSum-AllinOne and HighestN-AllinOne methods were replaced by the AllSumUnique and HighestNUnique methods, which allow scores to be calculated at the subpart and item level, a feature that was not available in past versions. Second, the 334 and 3345 methods have been replaced with the UniqueMaximums method. Finally, POES 2.0.1 incorporates three new scoring methods (CountWords, CountCat, and CountFreq). The purpose of this study is to compare the validity of hand scoring and the eight POES 2.0.1 scoring methods of the LEAS. To do so, we correlated the LEAS scoring methods with the Mayer-Salovey-Caruso Emotional Intelligence Test (MSCEIT; Mayer, Salovey, Caruso & Sitarenios, 2003).

## METHOD

### Participants

A total of 356 university students (204 females, 152 males) participated in this study in return for course credit. They ranged from 18 to 50 years (mean 19.95, SD 3.53). Participants identified themselves as follows: Caucasian (57.9%), Hispanic (12.4%), Asian (12.4%), African American (7.0%), Pacific Islander (6.2%), Native American (0.3%), and other (3.9%).

### Measures

The Mayer-Salovey-Caruso Emotional Intelligence Test (MSCEIT; Mayer, Salovey, Caruso & Sitarenios, 2003) is a 141-item test.

The Levels of Emotional Awareness Scale (LEAS; Lane et al. 1990) consists of 20 items that present emotionally-evocative scenarios. Each item has two subparts. The first subpart is the answer to the question, "How would you feel?" The second subpart is the answer to the question, "How would the other person feel?" The LEAS was scored both by hand or by POES computer scoring.



## METHOD

### Hand scoring the LEAS

LEAS scorers first search the two subparts for emotion words. Words describing thoughts rather than feelings (e.g., "confused") are scored 0. Words describing physical sensations (e.g., "dizzy") are scored 1. Words with positive or negative connotations, personality traits, or actions that are part of an emotional response (e.g., "rude," "honorable," "grudge," respectively) are scored 2. Finally, words identifying specific emotions (e.g., "happy") are scored 3. Next, scorers calculate Self and Other scores as the maximum value for words attributed to that person. If two separate, non-synonymous level 3 emotion words are given, the Self or Other score is raised to 4. Third, the scorer calculates the item score as the maximum of the Self and Other scores. When both Self and Other scores are 4 and the responses are not the same as each other, the item score is raised to 5. To calculate total scores across all items, the scores for the 20 items are summed.

### POES Computer Scoring the LEAS

Computer scoring was done using POES 2.0.1 (Ermini Leaf & Barchard, 2013). It uses eight scoring methods, most of which require a wordlist containing key words and phrases and their values. For example, the key word "happy" is scored 3. This study used LEAS Wordlist 2.5 (Barchard, 2013).

**CountWords.** The CountWords method counts the number of words in a response. This score can be calculated at the level of the subpart, the item, or the subject. We used subject-level scores.

**AllSum.** The AllSum method calculates the sum of all key values. First, POES searches each response for key words and phrases from the Wordlist. It notes the values of each of those keys. To calculate the AllSum score for a response, POES sums the values of the keys in that response. POES calculates AllSum scores for each subpart, item, and subject. We used subject-level scores.

**AllSumUnique.** The AllSumUnique method is based on the AllSum method but it eliminates repetitions from being scored twice. For example, imagine a response says, "I feel happy. I feel so happy I cry." To score this response, POES creates a *unique keys list*, which contains each key appearing in the response, but it only contains each key once. In the example, this list would contain the keys "happy" and "cry". The unique keys list also contains the values of each key. The AllSumUnique score is calculated as the sum of the values for each key in that list. AllSumUnique can be calculated at the level of subparts, items, or subjects. We used subject-level scores, so that if subjects use the same word in more than one response, they only receive credit once.

**HighestN.** The HighestN score is calculated as the sum of the highest N values of the keys that appear in a response. HighestN scores can be calculated at the subpart, item, or subject level. The user can specify the value of N. When N is set to 4 (the default), the item-level scores are identical to the Highest-4 method in previous versions of POES (Leaf & Barchard, 2010). In this study, we used subject-level scores with N equal to 40 to reduce the influence of long responses on the scores.

**HighestNUnique.** To score a response using the HighestNUnique method, POES first creates a unique keys list. Then the HighestNUnique score is calculated as the sum of the highest N values of the keys on that list. HighestNUnique scores can be calculated at the subpart, item, or subject level. For this study, we used subject-level scores with N equal to 40, so that if a subject used the same word in more than one response, they only received credit once.

**UniqueMaximums.** The UniqueMaximums method requires the user to set maximum values for each scoring level. The user sets values for KeyMax, SubpartMax, ItemMax, and SubjectMax. The scoring starts at the subpart level. For each subpart, an initial score is calculated as KeyMax or the highest key value in the response, whichever is lower. If there are multiple distinct keys that have the highest possible value, then the subpart score is incremented by one for each additional key. For example, if the response said, "I feel happy, jealous, and angry all at the same time" and KeyMax is set to 3, then the initial subpart score is 5 because "happy," "jealous", and "angry" each have a key score of 3. The final subpart score is calculated as the minimum of this initial subpart score and SubpartMax. For example, if SubpartMax is set to 4, then the subpart score is 4.

Next, the item score is calculated. The initial item score is calculated as ItemMax or the highest subpart score, whichever is lower. This initial item score is incremented by one for each subpart that has the maximum score. For example, if the score for subpart 1 was 4 and the score for subpart 2 was 4, then the initial item score is 5. The final item score is calculated as the minimum of this initial item score and ItemMax.

Finally, the subject score is calculated. The initial subject score is calculated as SubjectMax or the highest item score, whichever is lower. This initial subject score is incremented by one for each item that has the maximum score. For example, if the score for item 1 was 5 and the score for item 2 was 5, then the initial subject score is 6. The final subject score is calculated as the minimum of this initial subject score and SubjectMax.

**CountCat.** The CountCat method counts the number of distinct values that occur in a response. For example, if a response said, "I feel so sad I would cry", then "sad" has a value of 3 and "cry" has a value of 2, so that two distinct values (2 and 3) occur in this response. The CountCat score would be 2. Thus, CountCat is treating each value as a category and is counting the number categories that are represented by the keyed responses. The CountCat score can be calculated at the level of the subpart, the item, or the subject. For this study, we calculated the total of the item-level scores.

**CatFreq.** The CatFreq method counts the number of times each value occurs in a response. CatFreq produces as many scores as there are values. For example, imagine a response said, "I would be worried about her. She would be unhappy and might be thinking about suicide." This response contains four keys: "worried" with a value of 3, "unhappy" with a value of 3, "thinking" with a value of 0, and "suicide" with a value of 2. CatFreq0 would equal 1 because there was one key with a value of 0. CatFreq1 would equal 0 because no level 2 keys were used. Similarly, CatFreq2 and CatFreq3 would both equal 1. The CatFreq score can be calculated at the level of the subpart, the item, or the subject. We used the subject-level scores.

### Procedures

Participants completed the MSCEIT and the LEAS as a part of a larger online study.

## RESULTS

All LEAS scoring methods had significant positive correlations with total scores on the MSCEIT (see Table 1), thus demonstrating their convergent validity. The HighestN and HighestNUnique methods (both with N = 40) had the highest correlations with the MSCEIT. Most methods had higher correlations than CountWords: The only exceptions were the count of the number of level 0 emotion words (CatFreq0) and the count of the number of level 1 emotion words (CatFreq1). The UniqueMaximums method (with KeyMax = 3, SubpartMax = 4, and ItemMax = 5) had a slightly higher correlation than the hand scoring method.

Table 1  
*Correlations of LEAS Scoring with the MSCEIT (Total Score)*

LEAS Score	Correlation
Hand Scoring	.33***
POES Scoring	
CountWords	.21***
AllSum	.29***
AllSumUnique	.36***
HighestN†	.39***
HighestNUnique†	.40***
UniqueMaximums††	.34***
CountCat	.32***
CatFreq0	.20***
CatFreq1	.15**
CatFreq2	.25***
CatFreq3	.31***

\*\*  $p < .01$ . \*\*\*  $p < .001$ . † N = 40.

†† KeyMax = 3, SubpartMax = 4, ItemMax = 5.

## DISCUSSION

All of the POES 2.0.1 scoring methods had significant positive correlations with the MSCEIT, thus reinforcing the validity of the LEAS as a measure of emotional awareness. The highest correlations were for the HighestN (N = 40) and HighestNUnique (N = 40) methods. These correlations were larger than the correlations for hand scoring, thus making them a viable alternative to the time-consuming hand-scoring method.

One of the new scoring methods was CatFreq. It counts how often respondents use words that receive scores of 0, 1, 2, and 3. As might be expected, the frequency of level 3 emotion words (happy, sad, angry, etc.) had a higher correlation with the MSCEIT than the frequency of emotion words that receive lower scores. It appears that level 3 emotion words are driving the correlation between the LEAS and the MSCEIT.

Another new method in POES 2.0.1 is the UniqueMaximums method. This method is a generalization and replacement of the previous 334 and 3345 methods that were the closest possible approximation to hand-scoring. In this study, UniqueMaximums was set up with KeyMax = 3, SubpartMax = 4, and ItemMax = 5, so that it is equivalent to the 3345 method. Like previous research (Barchard et al., 2010), this method had slightly higher validity ( $r = .34$ ) than the hand-scoring method it approximates ( $r = .33$ ).

Another innovation in this version of POES is the AllSumUnique and HighestNUnique scoring methods. These methods give credit for an emotion word only the first time it appears. AllSumUnique had a higher correlation ( $r = .36$ ) with the MSCEIT than AllSum did ( $r = .29$ ). Similarly, HighestNUnique had a higher correlation ( $r = .40$ ) with the MSCEIT than HighestN did ( $r = .39$ ), although here the difference was trivially small (.01). Thus, it appears that duplicate words add little to our knowledge of someone's emotional awareness. What is most important is whether a person can describe the differences between various emotional experiences.

The last new method in POES 2.0.1 is CountCat. This method counts the number of distinct categories that were used in the responses, and thus indexes the richness of the emotional descriptions. In this dataset, CountCat had relatively high validity ( $r = .32$ ), similar to hand scoring ( $r = .33$ ), but not as good as the HighestN and HighestNUnique methods. This method deserves further exploration, perhaps with revised categories that more precisely examine the breadth of one's descriptions.

In conclusion, this study has demonstrated the validity of computer scoring of the LEAS. Some POES methods had higher correlations with the MSCEIT than hand-scoring. Moreover, having a computer do the scoring allows psychologists to use scoring methods that would be unwieldy if attempted by hand (AllSumUnique, HighestNUnique, UniqueMaximums), and thus allows us to measure new aspects of emotional awareness. In particular, the elimination of duplicate words across the entire set of 20 items allows us to measure the breadth of emotion knowledge, and counting the number of unique emotion categories allows us to measure the richness of emotional descriptions.