

ORIGINAL ARTICLE

The time has come to stop rotations for the identification of structures in the Hamilton Depression Scale (HAM-D₁₇)

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Objective: To use principal component analysis (PCA) to test the hypothesis that the items of the Hamilton Depression Scale (HAM-D₁₇) have been selected to reflect depression disability, whereas some of the items are specific for sub-typing depression into typical vs. atypical depression.

Method: Our previous study using exploratory factor analysis on HAM-D₁₇ has been re-analyzed with PCA and the results have been compared to a dataset from another randomized prospective study.

Results: PCA showed that the first principal component was a general factor covering depression disability with factor loadings very similar to those obtained in the STAR*D study. The second principal component was a bi-directional factor contrasting typical vs. atypical depression symptoms. Varimax rotation gave no new insight into the factor structure of HAM-D₁₇.

Conclusion: With scales like the HAM-D₁₇, it is very important to make a proper clinical interpretation of the PCA before attempting any form of exploratory factor analysis. For the HAM-D₁₇, our results indicate that profile scores are needed because the total score of all 17 items in the HAM-D₁₇ does not give sufficient information.

Keywords: Hamilton Depression Scale; principal component analysis; exploratory factor analysis

Introduction

Factor analysis refers to a variety of psychometric techniques employed with the objective of reducing the original universe of items in a rating scale to a smaller number of components or factors without losing the information stored in the individual items. If the researchers have no clear hypothesis as to how many factors there are for a given rating scale, factor analysis might be used as a means of exploring the data for a small set of factors. This form of factor analysis is called exploratory factor analysis.¹ If, however, factor analysis is used to test some hypotheses or expectations about factors and their clinical nature, it is then called confirmatory factor analysis. In this case, the number of factors and their loadings are tested on the same data, i.e., within the frame of the investigation under examination.²

The principal component analysis (PCA) method, which is essentially not a factor analysis,³ is often considered as an initial stage in factor analysis to identify by eigenvalues the number of factors to be considered in an exploratory factor analysis.⁴ However, as stressed by Hotelling,⁵ if we have a hypothesis based on the principle on which the items of the scale have been constructed, then PCA can be considered sufficient for testing this

hypothesis. Thus, when constructing intelligence tests (ability scales) or depression scales (disability scales), the items are carefully selected to be more or less positively correlated, resulting in the expectation that PCA will identify the first principal component as a general factor of intelligence (ability) or depression (disability). Hamilton actually selected the 17 items in his depression scale (HAM-D₁₇) to capture general depression disability.⁶ In intelligence tests, the second principal component is expected to contrast verbal vs. non-verbal intelligence by a bi-directional factor (positive vs. negative loadings), whereas in the HAM-D₁₇, it is expected to contrast typical vs. atypical depression.⁶⁻⁸ However, the literature on the HAM-D₁₇ is very unclear on studies using factor analysis, because different authors have used different techniques within their exploratory factor analyses, i.e., different forms of rotation.⁹ In the study by Fleck et al.,¹⁰ PCA identified a general factor in the HAM-D₁₇. Such a general factor is often considered as an argument for the total score being a sufficient statistic measure (unidimensionality), e.g., in Lewis et al.¹¹ However, the individual items of the HAM-D₁₇ within this general factor have very different loadings, implying that more than one dimension is present. Therefore, Fleck et al.¹⁰ then performed varimax rotation of the dataset in an attempt to identify other factors in an exploratory analysis in accordance with Kim & Mueller.¹ In this report, we will show the full outcome of PCA in the study by Fleck et al.¹⁰ as an example of using this method to test the hypothesis that the HAM-D₁₇ contains the general factor of depression disability as well as a second principal component

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separating the wings or subscale factors within the A,B,C version of HAM-D₁₇,⁷ in which (A) refers to HAM-D₆ (the core items of depression), (B) refers to the HAM-D₉ (the unspecific arousal symptoms of depression), and (C) refers to HAM-D₂ (suicidal thoughts and insight). These three subscales (A,B,C) have been selected on a purely clinical basis, not by factor analysis.⁷

Methods

Patients

The patients were all admitted to Psychiatric University Hospitals in Paris for depressive illness and were diagnosed according to DSM-III-R¹² using the Composite International Diagnostic Interview (CIDI), version 1.0.¹³ Exclusion criteria were as follows: serious medical disorder, organic mental disorder according to DSM-III-R, substance or alcohol disorders according to DSM-III-R, schizophrenia according to DSM-III-R, speech or hearing problems, or intelligence defect, i.e., an IQ of 70 or less.

The Hamilton Depression Scale (HAM-D₁₇)

The interview guide for HAM-D₁₇¹⁴ was used. All patients were assessed by the same interviewer (MFAF). The patients were all interviewed for HAM-D₁₇ within the first 3 days of hospitalization. At the end of this initial period, all patients were then interviewed with CIDI to arrive at a DSM-III-R diagnosis.

Statistical analysis

In the first publication of this study,¹⁰ PCA was directly connected with an exploratory factor analysis using varimax rotation. In the report presented herein, we performed a PCA in accordance with Hotelling,^{3,5} Duntman,¹⁵ and Child.⁴ We considered PCA as a purely mathematical analysis. When two items are correlated, the variance of each can be divided into two parts, one of which is common to both items (i.e., the general disability of depressive states) while the other is specific to each item and independent (i.e., orthogonal) of the common variance and the other specific variance. This independent, second component captures the specific variance, resulting in a bi-directional factor which contrasts by its negative vs. positive loadings.

When interpreting the symptom pattern of the second factor, we followed Child⁴ in considering all loadings (not only loadings with "statistical significance"), because they help capture the "flavor" of the factor loadings, emphasizing that PCA is a mathematical rather than a statistical model.

The PCA results from this report have been compared with the three-factor varimax factor analysis done by Fleck et al.,¹⁰ and with the post-hoc PCA analysis of STAR*D study results.⁶

Results

In total, 60 patients fulfilled the inclusion vs. exclusion criteria. The mean (SD) HAM-D₁₇ score was 26.6 (7.3).

The mean (SD) age was 47.0 (13.2) years. Females comprised 77% of the sample.

Six components with an eigenvalue of 1 or more were identified by PCA. The first principal component was a general factor (Table 1) with an eigenvalue of 4.16. The second principal component was a bi-directional factor (Table 2) with an eigenvalue of 2.22. Together, these two components explained 37.5% of the variance.

Table 1 shows the factor loadings of the first principal component. In both the present study and the STAR*D study, all HAM-D₁₇ items apart from insight had positive loadings, implying that the first principal component is a general factor of depressive disability.

Table 2 shows the negative and positive loadings for the second principal component. For comparison, the results of the three-factor varimax exploratory factor analysis published by Fleck et al.¹⁰ are also shown in Table 2. The negatively loaded items in the present study had 100% concordance with the factor 1 identified by Fleck et al.,¹⁰ including seven items, of which five are contained in the HAM-D₆ factor of specific depression symptoms (depressed mood, guilt, suicide, work and interests, motor retardation, and tiredness [general somatic]). The remaining item of psychic anxiety was actually identified by Fleck et al.¹⁶ in the Brazilian part of their study. The item of suicidal thoughts had the lowest loading among the negatively loaded items in Table 2. The positively loaded items cover factor 2 and factor 3 in the varimax rotation. Here, the insight item had the lowest loading.

Discussion

To the best of our knowledge, this re-analysis of the study by Fleck et al.¹⁰ is the first investigation to show the importance of clinically interpreting the results of a PCA before moving on to the various forms of factor analysis, as recommended by Kline.¹⁷

In a symposium on psychological factor analysis, Peel¹⁸ made an attempt to consider PCA as a method

Table 1 Principal components analysis: first principal component

		Fleck et al. ¹⁰	Bech et al. ⁶
1	Depressed mood	0.71	0.58
2	Guilt	0.33	0.29
3	Suicidal thoughts	0.20	0.36
4	Sleep initial	0.53	0.36
5	Sleep middle	0.62	0.33
6	Sleep delayed	0.53	0.32
7	Work and interests	0.53	0.42
8	Motor retardation	0.69	0.33
9	Agitation	0.54	0.26
10	Psychic anxiety	0.60	0.52
11	Somatic anxiety	0.57	0.56
12	Appetite decreased	0.42	0.48
13	Tiredness	0.52	0.32
14	Sexual interest	0.25	0.30
15	Hypochondriasis	0.23	0.31
16	Weight loss	0.52	0.40
17	Insight	-0.11	-0.10

Table 2 Principal components analysis with the second principal component and varimax rotation in the Fleck et al.¹⁰ study (items are listed according to their number in the Williams¹⁴ guide)

Fleck et al. ¹⁰ dataset			
Bi-directional second principal component		Varimax factor 1	
Negative loadings		+ = correspondence with PCA	
1	Depressed mood	-0.39	+
2	Guilt	-0.52	+
3	Suicide	-0.21	+
7	Work and interests	-0.46	+
8	Motor retardation	-0.34	+
13	General somatic	-0.42	+
14	Sexual interest	-0.37	+
PCA second component (present study)		Varimax factor 2	Varimax factor 3
Positive loadings			
4	Sleep initial	0.08	+
5	Sleep middle	0.00	+
6	Sleep late	0.20	+
9	Agitation	0.47	+
10	Psychic anxiety	0.48	+
11	Somatic anxiety	0.48	+
12	Appetite decreased	0.14	+
15	Hypochondriasis	0.55	+
16	Weight loss	0.25	+
17	Insight	0.00	+

to test hypotheses within intelligence tests and personality inventories with reference to a general principal component reflecting ability (general factor of intelligence) or disability (general factor of neuroticism), as well as to a specific principal component of ability (verbal vs. non-verbal intelligence) or disability (extraversion vs. introversion). As regards the two-component personality hypothesis, Peel¹⁸ referred to Eysenck's Personality Inventory. In 1969, Eysenck et al.¹⁹ concluded on their work with the Eysenck Personality Inventory: "A principal component solution on the two central factors (neuroticism and extraversion/introversion) gives a perfectly adequate approximation; a varimax rotation of the first two factors extracted may or may not improve this approximation". Therefore, Eysenck et al.¹⁹ preferred PCA results from the higher-order factor. This is most clearly discussed by Child,⁴ who states that the decrease of explained variance within PCA from the first to the last component and the simultaneous increase of error variance as one progresses from the first to the last component implies strongly that only the first and the second principal component are the objects for clinical interpretations.

Within the three-fold A,B,C HAM-D₁₇ version,^{7,8} which has been developed on a purely clinical basis, the HAM-D₉ or (B) contains the unspecific arousal symptoms which are to a large extent covered by the positively loaded items in the second PCA component (Table 2). In the varimax solution (Table 2), factor 3 covers sleep items, whereas factor 2 covers the remaining unspecific arousal items (HAM-D₉) or (B). The insight item had the lowest loadings in both the varimax solution and the PCA solution. The C wing of the HAM-D₁₇ includes the two items with the lowest loadings among both the negatively loaded items (suicide) and the positively loaded items

(insight). These two items (HAM-D₂) contain the most idiographic HAM-D items.^{7,8}

With our analysis of the HAM-D₁₇ dataset from the study by Fleck et al.,¹⁰ we have shown that a varimax rotation did not improve the approximation of identifying a general factor covering general depressive disability and a bi-directional factor covering typical vs. atypical depression. This finding is in concordance with the PCA of the STAR*D dataset in which the same version of the HAM-D₁₇ was used.⁶

As discussed by Kline,¹⁷ the demonstration of a general factor of disability is tautological, because it is a simple consequence of how Hamilton selected the items in the HAM-D₁₇, but should not be considered as an artefact of the PCA algebra. The clinical importance of PCA is the identification of the second principal component in which the items with negative loadings reflect the specific content of depression, whereas items with positive loadings reflect the unspecific symptoms of depression. The items not so clearly loaded (suicide and insight) are the idiographic (HAM-D₂) items.

As discussed by Salum et al.,²⁰ we need to use the typical HAM-D₁₇ items when measuring the antidepressive effects of drugs in the treatment of mild to moderate depression. The HAM-D₆ items are among the seven items Santen et al.²¹ found able to discriminate between paroxetine and placebo.

In conclusion, our analysis of the Fleck et al.¹⁰ dataset has shown that rotated factors can be seen as an artefact of factor analysis by changing the pattern of loadings already found clinically meaningful within the PCA approach. Our results thus indicate that profile scores are needed because the sum total score of all 17 items in the HAM-D₁₇ does not provide sufficient information about the structure of depression symptomatology.

Disclosure

The authors report no conflicts of interest.

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