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# Latent Profile Analysis of Cyberchondria and its Relationship with Health Information Overload and Avoidance

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# Introduction & Literature Review

- **Cyberchondria**

- Cyberchondria is a multidimensional, syndrome-like condition characterized by responses to anxiety, compulsivity, and associated negative emotional states (Starcevic et al., 2020).
- A **combination** of **behavioral patterns** and **concurrent emotional states** (McElroy et al., 2019).

- **Costs**

- A lifetime prevalence of 5.7% for health anxiety/hypochondriasis (Hannah et al., 2023).
- Cyberchondria severity was a significant predictor of the utilization (Kobryn & Duplaga, 2024).

**Table 3** Ordinal regression modeling of variables reflecting the utilization of healthcare services and alternative medicine

| Variable                        | Variable category | Visits to family physicians<br>OR (95%CI) | Visits to specialists<br>OR (95%CI) | Diagnostic procedures<br>OR (95%CI) | Hospital admission<br>OR (95%CI) | Emergency services<br>OR (95%CI) | Alternative medicine<br>OR (95%CI) |
|---------------------------------|-------------------|---|-------------------------------------|-------------------------------------|----------------------------------|----------------------------------|------------------------------------|
| Age                             |                   | 1.00 (0.99-1.01)                          | 0.99 (0.985-0.998)                  | 1.02 (1.01-1.02)                    | 0.98 (0.98-0.99)                 | 0.98 (0.97-0.99)                 | 0.99 (0.98-1.00)                   |
| CSS score                       |                   | 1.01 (1.001-1.012)                        | 1.01 (1.00-1.01)                    | 1.00 (1.00-1.017)                   | 1.01 (1.01-1.02)                 | 1.02 (1.01-1.02)                 | 1.02 (1.01-1.03)                   |
| SHAI_IL (illness likelihood)    |                   | 1.07 (1.05-1.09)                          | 1.05 (1.03-1.08)                    | 1.08 (1.06-1.10)                    | 1.04 (1.021-1.06)                | 1.04 (1.01-1.06)                 | 1.02 (0.99-1.04)                   |
| SHAI-NC (negative consequences) |                   | 0.89 (0.85-0.94)                          | 0.94 (0.90-0.98)                    | 0.92 (0.88-0.97)                    | 0.98 (0.93-1.03)                 | 0.97 (0.92-1.02)                 | 1.03 (0.97-1.10)                   |

# Introduction & Literature Review

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- **Dimensions of Cyberchondria**

- Cyberchondria may be better defined as a **dimensional construct**, on a continuum from a mild to severe behavioural and psychopathological pattern (Starcevic et al., 2020).
- **Four Dimensions** (McElroy & Shevlin, 2014):
  - **Compulsion:** The degree to which internet searches for health information interfere with daily activities.
  - **Distress:** The tendency to experience anxiety while searching for health information online.
  - **Excessiveness:** The repetitive nature of the searches.
  - **Reassurance:** The need to seek reassurance from medical professionals or online sources.

- **Theories and Model**

- Reassurance-Seeking Model (Doherty-Torstrick et al., 2016)
- Metacognition of Health Anxiety (Airoldi et al., 2022; Fergus & Spada, 2018)

# Introduction & Literature Review

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- **Variable-Centered or Person-Centered Approach ?**
  - **Variable-Centered Approach** (McElroy & Shevlin, 2014; Jungmann & Witthöft, 2020)
    - Studying the relationships between variables across a population.
  - **Person-Centered Approach**
    - Instead of looking at the average effect of variables, it looks at how individuals with different profiles or patterns within the data might respond differently.
- *Research Question 1: What latent profiles can be identified among individuals exhibiting diverse patterns of cyberchondria?*

# Introduction & Literature Review

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- **Health Information Overload, Avoidance and Cyberchondria**
  - **Health Information Overload**
    - Excessive intake of health information becomes a burden rather than a resource (Bawden & Robinson, 2015).
    - **Reassurance-Seeking Model:** Individuals with high health anxiety engage in excessive reassurance-seeking, which fails to alleviate anxiety, perpetuating distress (Doherty-Torstrick et al., 2016).
    - **Social Amplification of Risk Framework (SARF):** Heightened exposure to health-related risks amplifies perceived risks (Li et al., 2023), contributing to health anxiety and cyberchondria.

# Introduction & Literature Review

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- **Health Information Overload, Avoidance and Cyberchondria**
  - **Health Information Avoidance**
    - behavior intended to prevent or delay the acquisition of available but potentially unwanted information (Howell & Shepperd, 2013).
    - **A potential strategy?**
      - A coping mechanism to mitigate the negative effects of cyberchondria (Doherty-Torstrick et al., 2016; Laato et al., 2020).
      - Avoidance may help individuals with high health anxiety avoid further distress from online searches.
      - The complete avoidance of health information is unrealistic in an era where the Internet is a dominant source of medical knowledge (Starcevic et al., 2020).
      - Information avoidance can temporarily alleviate health anxiety, it is insufficient without concurrent efforts to improve health literacy and online information competency (Luo, 2019).
- *Research Question 2: What is the relationship between health information overload, avoidance, and the subtypes of cyberchondria?*

# Methodology

- **Participants**

- Data was collected through a questionnaire, with participant recruitment facilitated by the Credamo ([www.credamo.com](http://www.credamo.com)), a member of the European Society for Opinion and Marketing Research (ESOMAR), which is recognized for adhering to international authoritative standards.
- Data collection conducted from May 13th to May 17th, 2024.
- 500 respondents (65.8% female,  $N = 329$ ;  $M_{\text{age}} = 30.16$ ,  $SD = 6.78$ ).

Table 2. Demographic Variables Descriptive Statistics

| Variables                              | <i>M (SD) or N (%)</i> |
|--|------------------------|
| Age                                    | 30.16 (6.79)           |
| Gender                                 |                        |
| -Female                                | 329 (65.8)             |
| -Male                                  | 171 (34.2)             |
| Hukou                                  |                        |
| -Urban resident household              | 335 (67.0)             |
| -Rural resident household              | 165 (33.0)             |
| Educational Level                      |                        |
| -Technical school/High school or below | 6 (1.2)                |
| -Associate degree                      | 28 (5.6)               |
| -Bachelor's degree                     | 377 (75.4)             |
| -Graduate degree                       | 89 (17.8)              |
| Monthly Income                         |                        |
| -1000 RMB or less                      | 8 (1.6%)               |
| -1001-3000 RMB                         | 73 (14.6%)             |
| -3001-5000 RMB                         | 63 (12.6%)             |
| -5001-8000 RMB                         | 141 (28.2%)            |
| -8001-10000 RMB                        | 98 (19.6%)             |
| -More than 10000 RMB                   | 117 (23.4%)            |

# Methodology

- **Measurements**

- **Cyberchondria**

- Cyberchondria Severity Scale (CSS-12; McElroy et al., 2019);
    - 5-point Likert scale ( $M \pm SD = 3.63 \pm 0.63$ , Cronbach's  $\alpha = 0.827$ )

- **Health Information Overload and Avoidance**

- Health information overload was measured using three items adapted from Song et al. (2021); Health information avoidance was assessed with four items from Howell & Shepperd's (2016) research.
    - 5-point Likert scale ( $M_{overload} \pm SD_{overload} = 2.87 \pm 1.17$ , Cronbach's  $\alpha_{overload} = 0.807$ ;  $M_{avoidance} \pm SD_{avoidance} = 2.36 \pm 0.90$ , Cronbach's  $\alpha_{avoidance} = 0.882$ ).

Table 3. Constructs Measurements

| Constructs  | Items  | Cronbach's $\alpha$ |
|---|--|---------------------|
| Health Information Overload<br>(Song et al., 2021)        | 1 There was too much health information from media so that I was burdened in handling it.  | 0.807               |
|   | 2 I could not effectively handle all the health information from the media.  |                     |
|   | 3 Because of the plenty health information from the media, I felt it difficult to acquire all the information.   |                     |
| Health Information Avoidance<br>(Howell & Shepperd, 2016) | 1 There is some health information I would rather not know.  | 0.882               |
|   | 2 I think ignorance is bliss with respect to certain health information.   |                     |
|   | 3 I avoid obtaining health information.  |                     |
|   | 4 I can think of situations where I would avoid health information.  |                     |
| Cyberchondria<br>(McElroy et al., 2019)                   | 1 If I notice an unexplained bodily sensation I will search for it on the internet.  | 0.827               |
|   | 2 Researching symptoms or perceived medical conditions online distracts me from reading news/sports/entertainment articles online.                     |                     |
|   | 3 I read different web pages about the same perceived condition.   |                     |
|   | 4 I start to panic when I read online that a symptom I have is found in a rare/serious condition.  |                     |
|   | 5 Researching symptoms or perceived medical conditions online leads me to consult with my GP.  |                     |
|   | 6 I enter the same symptoms into a web search on more than one occasion.   |                     |
|   | 7 Researching symptoms or perceived medical conditions online interrupts my work (e.g., writing emails, working on word documents or spreadsheets).    |                     |
|   | 8 I think I am fine until I read about a serious condition online.   |                     |
|   | 9 I feel more anxious or distressed after researching symptoms or perceived medical conditions online.   |                     |
|   | 10 Researching symptoms or perceived medical conditions online interrupts my offline social activities (e.g., reduces time spent with friends/family). |                     |
|   | 11 I suggest to my GP/medical professional that I may need a diagnostic procedure that I read about online (e.g., a biopsy/a specific blood test).     |                     |
|   | 12 Researching symptoms or perceived medical conditions online leads me to consult with other medical specialists (e.g., consultants).                 |                     |



# Data Analysis & Results

- **Latent Profile Analysis of Cyberchondria**

- R version 4.3.1, tidyLPA package
- **Akaike Information Criterion (AIC):** Lower values indicate a better model fit.
- **Bayesian Information Criterion (BIC):** Lower values indicate a better model fit.
- **Log-Likelihood (LL):** Lower values indicate a better model fit.
- **Bootstrap Likelihood Ratio Test (BLRT):** A  $p$ -value  $< 0.05$  suggests that the  $k$ -class model provides a statistically superior fit compared to the  $k-1$  class model.
- **Entropy:** Values  $\geq 0.8$  indicate over 90% classification accuracy.

**Table 4.** Results of model fitting for potential latent analysis of cyberchondria

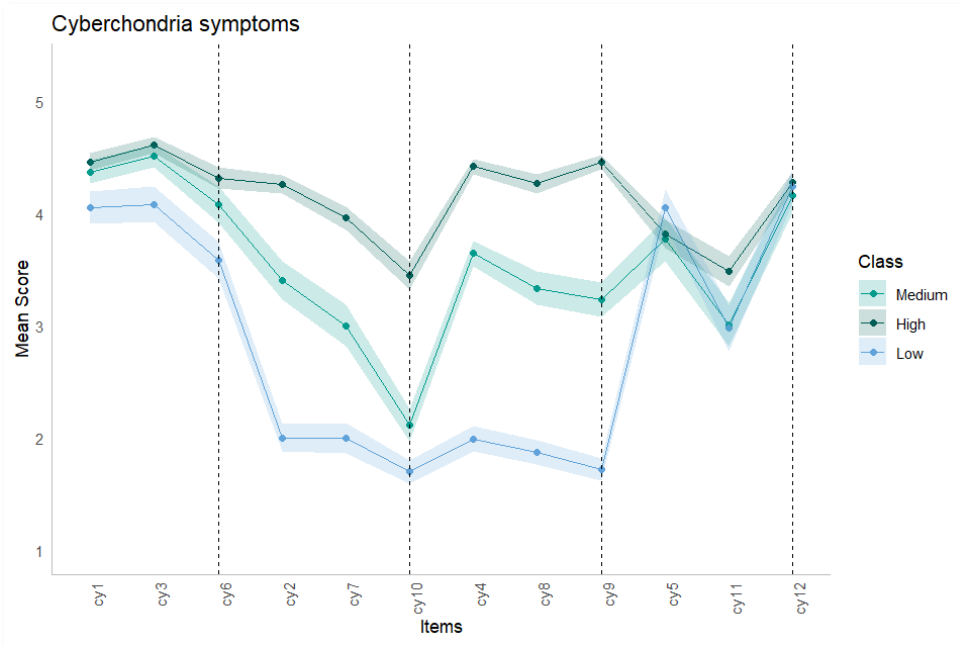
|                    | 1 class   | 2 class   | 3 class   | 4 class   | 5 class   |
|--------------------|-----------|-----------|-----------|-----------|-----------|
| Log-likelihood     | - 8721.37 | - 7724.25 | - 7566.97 | - 7366.35 | - 7315.74 |
| AIC                | 17490.74  | 15522.50  | 15233.94  | 14858.71  | 14783.49  |
| BIC                | 17811.04  | 15678.44  | 15444.67  | 15124.23  | 15103.80  |
| BLRT               | N/A       | 1994.24   | 314.56    | 401.24    | 101.22    |
| p-value for BLRT   | N/A       | $< 0.01$  | $< 0.01$  | $< 0.01$  | $< 0.01$  |
| Entropy            |           | 0.96      | 0.88      | 0.90      | 0.91      |
| Smallest Class (%) |           | 34 %      | 23 %      | 3 %       | 3 %       |

*Note.* AIC=Akaike Information Criterion, BIC=Bayesian Information Criterion, BLRT=Bootstrapped Likelihood Ratio Tests.

# Data Analysis & Results

## • Latent Profile Analysis of Cyberchondria

*Research Question: What latent profiles can be identified among individuals exhibiting diverse patterns of cyberchondria?*



- **Medium Compulsion and Distress Group**
  - 22.6% of the sample ( $N = 113$ )
  - Scores around the mean on compulsion and distress dimensions
- **High Compulsion and Distress Group**
  - 48.2% of the sample ( $N = 241$ )
  - Higher-than-average scores on both compulsion and distress
- **Low Compulsion and Distress Group**
  - 29.2% of the sample ( $N = 146$ )
  - lower scores on excessiveness, compulsion, and distress dimensions

**Table 5.** Descriptive statistics and test of difference for each latent category of cyberchondria on different dimensions

| Cyberchondria symptoms | Total ( $N = 500$ ) | Class 1 ( $N = 113$ ) | Class 2 ( $N = 241$ ) | Class 3 ( $N = 146$ ) | F-value | p-value |
|------------------------|---------------------|-----------------------|-----------------------|-----------------------|---------|---------|
| Excessiveness          | 4.27 ± 0.63         | 4.32 ± 0.45           | 4.46 ± 0.46           | 3.90 ± 0.81           | 36.90   | < 0.001 |
| Compulsion             | 3.07 ± 1.02         | 2.84 ± 0.56           | 3.89 ± 0.54           | 1.90 ± 0.54           | 94.03   | < 0.001 |
| Distress               | 3.42 ± 1.15         | 3.40 ± 0.44           | 4.38 ± 0.38           | 1.86 ± 0.43           | 214.2   | < 0.001 |
| Reassurance            | 3.78 ± 0.74         | 3.65 ± 0.73           | 3.86 ± 0.66           | 3.76 ± 0.86           | 0.98    | 0.321   |

# Data Analysis & Results

- **The relationship between health information overload, avoidance and subtypes of cyberchondria**

- Logistic Regression, R version 4.3.1, nnet package
- **Gender Differences:**
  - Females are more likely than males to belong to the medium compulsion and distress group (OR = 0.46,  $p < 0.01$ ).
- **Hukou Status:**
  - Urban residents are less likely than rural residents to belong to the medium (OR = 0.48,  $p < 0.05$ ) or high compulsion and distress group (OR = 0.42,  $p < 0.05$ ).
- **Health Information Overload:**
  - Participants with higher levels of information overload are significantly more likely to be in the medium (OR = 2.14,  $p < 0.001$ ) or high compulsion and distress group (OR = 3.60,  $p < 0.001$ ).
- **Health Information Avoidance:**
  - Avoidance behaviors do not significantly impact classification into the medium (OR = 1.32,  $p > 0.05$ ) or high compulsion and distress group (OR = 1.35,  $p > 0.05$ ).

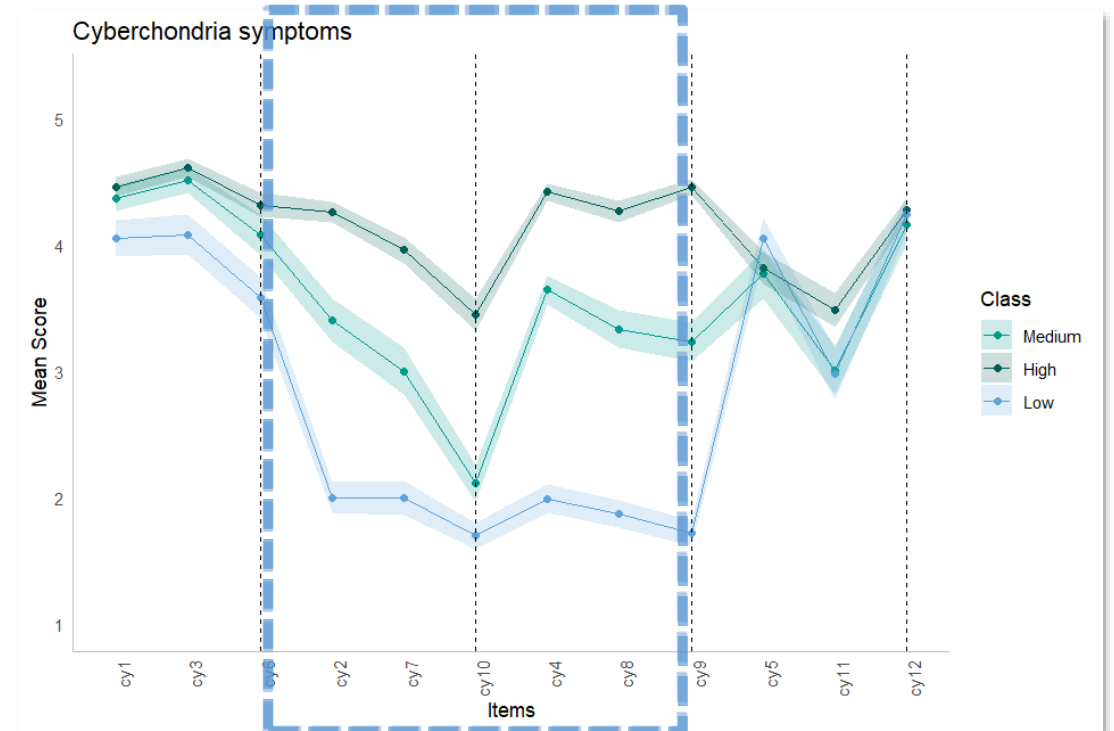
**Table 6.** Results of multinomial logistic regressions on the three latent category groups

|                              | Group with medium compulsion and distress |      |          |              | Group with high compulsion and distress |      |          |              |
|------------------------------|---|------|----------|--------------|---|------|----------|--------------|
|                              | <i>b</i> ( <i>SE</i> )                    | OR   | <i>p</i> | 95% CI       | <i>b</i> ( <i>SE</i> )                  | OR   | <i>p</i> | 95% CI       |
| Age                          | -.00 (.02)                                | 1.00 | .94      | [.95, 1.05]  | -.01 (.02)                              | .99  | .60      | [.95, 1.03]  |
| Gender                       | -.77 (.30)                                | .46  | < .01    | [.26, .84]   | -.31 (.27)                              | .73  | .26      | [.43, 1.26]  |
| Highest education degree     | -.09 (.28)                                | .92  | .76      | [.53, 1.59]  | -.09 (.26)                              | .91  | .73      | [.54, 1.53]  |
| Monthly salary               | -.22 (.13)                                | .80  | .10      | [.62, 1.04]  | -.10 (.13)                              | .91  | .44      | [.71, 1.16]  |
| Hukou                        | -.74 (.34)                                | .48  | < .05    | [.24, .93]   | -.86 (.33)                              | .42  | < .05    | [.22, .80]   |
| Health information overload  | .76 (.18)                                 | 2.14 | < .001   | [1.52, 3.02] | 1.28 (.17)                              | 3.60 | < .001   | [2.60, 4.99] |
| Health information avoidance | .28 (.20)                                 | 1.32 | .17      | [.89, 1.96]  | .30 (.19)                               | 1.35 | .11      | [.93, 1.95]  |

Note. reference group for multinomial logistic regression is group with Low Compulsion and Distress'; reference category for gender is 'female'; for highest education degree is 'primary school and below'; for monthly salary is '1000 yuan and below'; for hukou is 'rural population'.

# Conclusion & Discussion

- **Heterogeneity within cyberchondria**
  - Different combinations of psychological dimensions.
  - Personalized Interventions
- **Key Intervention Targets**
  - **Compulsion**
    - Reduce excessive health-related search behaviors.
    - Provide health information management skills and guide users toward balanced information-seeking behaviors.
  - **Distress**
    - Alleviate psychological stress.
    - Implement emotional management and psychological support interventions, such as Cognitive Behavioral Therapy (CBT) and relaxation training.



# Conclusion & Discussion

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- **Population Differences in Intervention Design**

- **Gender Differences**

- Women are more likely to belong to the medium Compulsion and Distress group.
- Associated with higher health-related anxiety and more frequent health information-seeking behavior (Lagoe & Atkin, 2015; WHO, n.d.).

- **Hukou Status Differences**

- Rural residents are more likely to belong to the medium or high Compulsion and Distress groups.
- Lower health literacy and socioeconomic factors contribute to higher health anxiety (Li et al., 2021).
- Another Dimension of Digital and Health Inequality
  - Access to the internet, but problematic usage.
  - The digital divide is not fully bridged.

# Conclusion & Discussion

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- **Role of Information Behavior**

- **Health Information Overload**

- A significant predictor for the medium and high Compulsion and Distress groups.
- Based on the Social Amplification of Risk Framework (SARF)
  - High exposure to risk-related information intensifies health anxiety and compulsive behaviors (Zheng et al., 2023; Li et al., 2023).

- **Health Information Avoidance**

- Not a significant predictor of cyberchondria subtypes.
- While avoidance may temporarily reduce anxiety, it is insufficient for preventing cyberchondria (Starcevic et al., 2020; Roth & Cohen, 1986).

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