

User Preference in Chinese E-Commerce Interface Design: A Semantic Differential Approach

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ABSTRACT

As consumer demands for user experience in e-commerce platforms continue to rise, interface layout has emerged as a critical factor influencing purchase decisions. This study investigates design preferences across four major Chinese platforms—JD.com (utilitarian layout), Taobao (social-commerce integration), Douyin Mall (short-video-driven interface), and Pinduoduo (gamified presentation). Employing a mixed-methods approach, we conducted semi-structured interviews ($n = 32$) and a semantic differential survey ($n = 227$ valid responses) to establish a perceptual evaluation system. Six bipolar adjective pairs (e.g., “orderly vs. cluttered,” “harmonious vs. conflicting”) were used to construct perceptual-semantic spaces. Principal component analysis (PCA) via SPSS 26.0 identified three core dimensions: Interface Layout ($\alpha = .87$), Information Presentation ($\alpha = .83$), and User Preference ($\alpha = .79$). Results demonstrated significant correlations between spatial zoning patterns and user satisfaction ($r = .68$, $p < .01$), with JD.com ranking highest due to its structured layout, while Douyin Mall scored lower owing to its video-centric immersion. These findings offer empirical guidelines for cross-cultural interface optimization and propose a triaxial design framework balancing layout, information, and emotional engagement.

Keywords: E-commerce interface design, Semantic differential method, User experience, Principal component analysis

INTRODUCTION

In contemporary society, users increasingly demand higher-quality experiences on e-commerce platforms, where the visual appeal of the primary interface directly impacts user satisfaction and loyalty. According to the 54th Statistical Report on China’s Internet Development (2024), China’s internet user base reached nearly 1.1 billion (1.09967 billion) by June 2024. Within China’s new media landscape, short-video platforms such as Douyin, Kuaishou, and TikTok have driven the rapid integration of new media and e-commerce through fragmented information dissemination and high interactivity. These platforms are characterized by dynamic content delivery, intuitive interfaces, and rapid information propagation. As market demands and user preferences evolve (Jin et al., 2023), designing homepage layouts that align with new media communication paradigms has become critical. In the current digital era, e-commerce platforms have developed

stable multipolar structures over time, with homepage design emerging as a pivotal factor influencing user experience (Chen et al., 2022).

In China, research has focused on analyzing strategies and models of new media. On one hand, enterprises aim to attract consumers through creative content, brand image cultivation, and user interaction on short-video platforms. On the other hand, studies emphasize the impact of interface design on viewers. Research indicates that hierarchical interface design can enhance brand awareness and reputation, thereby improving consumer affinity. Additional studies have explored educational, communicative, and content-driven short videos as distinct categories.

Internationally, scholars have also achieved significant advancements in new media interface design. Foreign studies prioritize investigating user behavior and psychology on short-video platforms to better understand the role of interface design. Beyond traditional quantitative metrics, international research increasingly explores qualitative measurements, such as user emotions and engagement levels (Chen et al., 2022). Optimization strategies in video design emphasize innovative content creation, perspectives (first- and third-person narratives), and collaborations with social media ecosystems. Furthermore, some studies address legal and ethical challenges in short-video marketing to ensure corporate compliance and social responsibility in campaigns (Gan et al., 2023).

Through the aforementioned research, the retrieved literature was analyzed using CiteSpace software (Geng et al., 2024), with a focus on keywords including “e-commerce platforms,” “semantic differential method,” and “interface design.” Following systematic searches, 211 qualified papers were classified, synthesized, and summarized. The analytical results revealed that under the themes of “new media” and “interface design,” high-frequency concepts such as “media convergence,” “interaction design,” “user experience,” and “mobile applications” predominated. These studies predominantly adopted user-centered perspectives, thereby substantiating the critical need for research on new media marketing platforms (Zhang et al., 2024).

EXPERIMENTAL PROCESS FOR USER PREFERENCES IN CHINESE E-COMMERCE INTERFACE DESIGN

The prerequisite for optimizing new media interface design lies in constructing a perceptual sample space. By establishing perceptual-semantic spaces for e-commerce platforms—comprising both sample spaces (platform selections) and semantic spaces (affective lexicons)—we conducted perceptual preference surveys through semantic mapping. The detailed experimental workflow is illustrated in Fig. 1.

New media interfaces play a pivotal role in user experience and commercial marketing. To systematically investigate the characteristics and effectiveness of new media interfaces, representative samples must be selected from the constructed sample space following the experimental workflow outlined earlier (Fu et al., 2023). Through literature review and consultations with 32 industry practitioners and experts, we identified the following categories of new media marketing platforms.

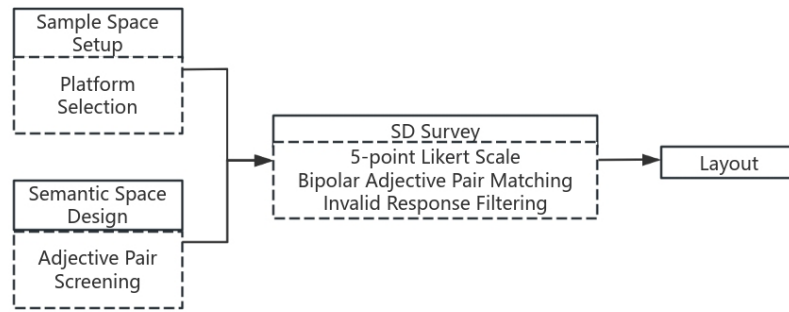


Figure 1: Experimental workflow diagram.

Table 1: Categories of new media marketing platforms.

	Representative Cases	Core Features
E-commerce Platforms (Product-Centric)	JD.com Suning.com Vipshop Pinduoduo	Facilitate product marketing through promotional campaigns (e.g., flash sales, coupons) and strategic advertising placements (banners, search rankings).
Life Service Platforms	Meituan Dianping Ele.me	Enable merchants to attract consumers via service packages (e.g., discounted combos) and targeted ad delivery based on geolocation/user preferences.
Content-Creation & Sharing Platforms	Bilibili Toutiao Zhihu	Support brand promotion through UGC/PGC content dissemination (videos, articles), Q&A engagement, and algorithm-driven ad exposure optimization.
Travel Service Platforms	Ctrip Qunar Fliggy	Aggregate tourism products (hotels, tours) with dynamic pricing strategies and immersive multimedia showcases (360° views, VR experiences).
Financial Service Platforms	Alipay JD Baitiao	Integrate financial products (loans, insurance) with behavioral targeting, leveraging user transaction data for personalized recommendations.
Enterprise Service Platforms	Alibaba Cloud Tencent Cloud	Provide B2B solutions (cloud computing, big data analytics) coupled with case study showcases and webinar-based technical marketing.

Continued

Table 1: Continued

	Representative Cases	Core Features
Online Education Platforms	NetEase Cloud Classroom, Tencent Classroom, TAL Education	Drive course enrollment through free trial sessions, user review systems, and AI-powered learning path recommendations.

Through comparative analysis of the four selected platforms, the study revealed that all platforms prioritize and meticulously design their app homepage interfaces. The visual layout of the homepage directly shapes users’ initial impressions during the first interaction, where effective interface design guides attention, stimulates exploratory intent, and ultimately enhances purchasing impulses, thereby significantly improving marketing outcomes. Based on these findings, this research selected JD.com (utilitarian layout), Taobao (social-commerce integration), Douyin Mall (short-video-driven interface), and Pinduoduo (gamified presentation) as representative cases of new media marketing platforms.

To eliminate confounding factors, this study adopted the design methodology of (Zou, 2023) by applying grayscale processing to all homepages and masking recommended products, retaining only original app icons and textual elements. Product images and prices were replaced with uniform color blocks. Additionally, screenshots were captured on new devices after logging out of user accounts to mitigate bias from personal preference (see Fig. 2).

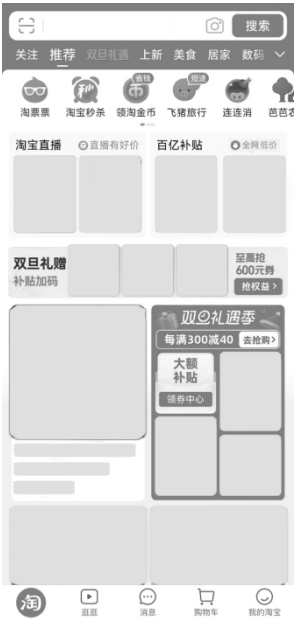


Figure 2: Sample space processing.

Development of Perceptual-Semantic Space and Lexicon

Refinement “Perceptual imagery” refers to human sensory responses to objects, and perceptual lexicon serves as linguistic representations of such responses (Hu et al., 2024). In new media marketing platforms, user perceptions of interface design are often implicit. To establish correlations between design elements and perceptual preferences, a curated set of affective adjective pairs must first be selected for semantic space construction. Through multi-source data collection—including interviews with marketers and users ($n = 32$), literature reviews, and online resources—120 affective terms were initially gathered and categorized into five domains: Visual Perception, Information Presentation, Interaction Experience, Emotional Atmosphere, and Brand Image (He et al., 2025).

Given semantic redundancies (e.g., synonymous pairs) and irrelevance to interface design, direct use of all 120 terms would impose cognitive overload on participants. Therefore, iterative screening was conducted: (1) Primary Screening: Merging synonyms and eliminating ambiguous terms, reducing the lexicon to 40 validated pairs; (2) Secondary Screening: Matching the 40 pairs with interface samples via questionnaire surveys, where participants selected the most representative terms (see Fig. 3).



Please select 10 sets from the following 40 descriptive word pairs according to your perception of the interface design of these four photos. 【请选择6项,已选择0项】

- | | | |
|---|--|--|
| <input type="checkbox"/> Luminous vs. Dim | <input type="checkbox"/> Refined vs. Coarse | <input type="checkbox"/> Sharp vs. Blurred |
| <input type="checkbox"/> Harmonious vs. Conflicting | <input type="checkbox"/> Comprehensive vs. Sparse | <input type="checkbox"/> Understandable vs. Abstruse |
| <input type="checkbox"/> Novel vs. Conventional | <input type="checkbox"/> Vibrant vs. Dull | <input type="checkbox"/> Trendy vs. Outdated |
| <input type="checkbox"/> Ornate vs. Plain | <input type="checkbox"/> Spacious vs. Narrow | <input type="checkbox"/> Accurate vs. Inaccurate |
| <input type="checkbox"/> Prominent vs. Hidden | <input type="checkbox"/> Effortless vs. Cumbersome | <input type="checkbox"/> Minimalist vs. Cluttered |
| <input type="checkbox"/> Spatially Balanced vs. Cramped | <input type="checkbox"/> Orderly vs. Chaotic | <input type="checkbox"/> Conspicuous vs. Subtle |
| <input type="checkbox"/> Detailed vs. Concise | <input type="checkbox"/> Complete vs. Incomplete | <input type="checkbox"/> Smooth vs. Laggy |

Figure 3: Questionnaire format for the secondary screening.

The questionnaire paired 40 affective adjective pairs with interface sample images. Participants were instructed to select 6 pairs that best matched each interface. Following preliminary screening and semantic merging of overlapping terms, Table 2 presents the final 6 validated bipolar adjective pairs, as follows:

Table 2: Perceptual semantic vocabulary pairs.

	Visual Perception	Information Presentation
Sensory Adjective Pairs	Orderly vs. Cluttered	Prominent vs. Hidden
	Harmonious vs. Conflicting	Complete vs. Incomplete
Affective Adjective Pairs	-	Novel vs. Conventional
	Likable vs. Dislikable	

The survey adopted a scale-based format, integrating the six validated affective adjective pairs with interface design samples. Participants evaluated four distinct platform interfaces using a 5-point Likert scale (−2 to +2). Demographic questions regarding age, education level, and monthly expenditure were included to contextualize subsequent analyses (Max Liang & Lian, 2021).

From the initial 231 responses, 4 were excluded due to incomplete or inconsistent answers, yielding 227 valid responses (valid rate: 98.27%). Demographic analysis revealed a balanced gender distribution (male: 41.85%, female: 58.15%), with the majority aged 18–25 years (87.2%), indicating familiarity with new media platforms. Participants were predominantly undergraduate and college students (91.6%), demonstrating adequate literacy for survey comprehension. Monthly expenditures averaged RMB 2,000–4,000 (≈USD 280–560), reflecting sufficient disposable income for online consumption.

FACTOR ANALYSIS OF PERCEPTUAL EVALUATIONS

The reliability analysis via SPSS yielded a Cronbach’s α of 0.965, exceeding the threshold of 0.8, which indicates high internal consistency of the scale. For validity, the KMO value was 0.905 (>0.8), confirming suitability for factor analysis, and Bartlett’s test of sphericity showed statistical significance ($p < 0.001$). All common factor loadings surpassed 0.7, demonstrating strong explanatory power of the extracted factors.

In factor analysis, the total variance explained reached 81.6% through the first three factors. Although no single factor achieved 80% cumulative variance, the first factor had the highest eigenvalue (3.82), explaining 45.1% of the variance, followed by the second (23.3%) and third (13.2%) factors. Thus, the first three factors were selected as principal components, representing the core dimensions of the six original variables.

Factor loadings revealed the following patterns:

- **Factor 1 (Information Clarity):** Dominated by *Orderly vs. Cluttered* (0.883) and *Harmonious vs. Conflicting* (0.638), emphasizing the importance of explicit information presentation.
- **Factor 2 (Innovation Preference):** Driven by *Prominent vs. Hidden* (0.686) and *Complete vs. Incomplete* (0.873), reflecting users' inclination toward innovative designs.
- **Factor 3 (Visual Harmony):** Characterized by *Novel vs. Conventional* (0.727) and *Harmonious vs. Conflicting* (0.874), indicating users' prioritization of visual coherence.

The findings categorize key factors influencing interface design of new media marketing platforms into three dimensions:

- **Interface Layout Factor (F_1):** Comprising *Orderly vs. Cluttered* and *Harmonious vs. Conflicting*, reflecting users' perception of visual order and harmony.
- **Information Presentation Factor (F_2):** Encompassing *Prominent vs. Hidden* and *Complete vs. Incomplete*, emphasizing clarity and completeness of information.
- **User Preference Factor (F_3):** Defined by *Novel vs. Conventional* and *Likable vs. Dislikable*, capturing users' inclination toward innovative designs and emotional preferences.

Based on the component score coefficient matrix, the factor score equations are:

$$F_1 = 0.765X_1 + 0.566X_2 - 0.445X_3 + 0.315X_4 - 0.323X_5 - 0.136X_6$$

$$F_2 = -0.361X_1 + 0.200X_2 + 0.621X_3 - 0.350X_4 - 0.233X_5 + 0.830X_6$$

$$F_3 = -0.118X_1 - 0.496X_2 + 0.222X_3 + 0.408X_4 + 1.024X_5 - 0.375X_6$$

where X_1 to X_6 represent the ratings of the six bipolar adjective pairs.

The composite score equation, weighted by total variance explained, is:

$$F_z = 0.2802F_1 + 0.55225F_2 + 0.8151F_3$$

This equation synthesizes the contributions of the three factors to quantify overall user preference for interface design.

Table 3: Component score coefficient matrix.

	1	2	3
Orderly vs. Cluttered	.765	-.361	-.118
Harmonious vs. Conflicting	.566	.200	-.496
Prominent vs. Hidden	-.445	.621	.222
Complete vs. Incomplete	.315	-.350	.408
Novel vs. Conventional	-.323	-.233	1.024
Likable vs. Dislikable	-.136	.830	-1.375

As shown in Table 3, higher composite scores (F_z) indicate stronger user preference for interfaces perceived as “complete,” “prominent,” “harmonious,” “orderly,” and “novel,” whereas lower scores correlate with negative perceptions of “incomplete,” “hidden,” “conflicting,” “cluttered,”

and “conventional” designs. Using the equations (F_1, F_2, F_3, F_z) and interface samples from JD.com, Taobao, Douyin Mall, and Pinduoduo, the composite perceptual scores were calculated and ranked in descending order:

Table 4: Sample human systems integration test parameters (Folds et al., 2008).

	F_1	F_2	F_3	F_z
JD.com	0.595	-0.028	0.028	0.172
Taobao	-0.64	0.148	-0.247	-0.3
Douyin Mall	0.422	-0.16	-0.598	-0.46
Pinduoduo	-0.19	-0.377	-0.707	-0.838

From the perspective of factor scores, the composite score (F_z) represents users’ holistic perception of interface design. A higher absolute F_z value indicates stronger polarization in evaluations (extreme preference or aversion). The data show JD.com with the highest F_z (0.172), followed by Taobao (-0.3) and Pinduoduo (-0.46), while Douyin Mall ranked lowest (-0.838).

The high User Preference scores of JD.com and Taobao ($F_3 > 1.0$) can be attributed to their balanced layout and information architecture, which reduce cognitive load and stimulate exploration and purchasing intent. In contrast, Douyin Mall’s homepage prioritizes livestream content, impairing product discoverability ($F_2 \downarrow$) and causing visual fatigue ($F_1 \downarrow$), thereby increasing user attrition risks.

JD.com & Pinduoduo: High Interface Layout scores ($F_1 > 0.4$) reflect user preference for their modular designs (categorical navigation bars, grid-based product displays).

JD.com & Taobao: Strong Information Presentation scores ($F_2 > -0.03$) stem from intuitive product detail pages and efficient search filters.

Douyin Mall: As the only platform featuring a livestream-driven homepage, its immersive video feed led to disorganized spatial zoning ($F_1 = -0.19$) and obscured key information ($F_2 = -0.377$), resulting in the lowest User Preference score ($F_z = -0.838$).

CONCLUSION

Based on the three key factors (F_1, F_2, F_3) extracted through the Semantic Differential Method, this study proposes the following design principles:

Guided by Factor F_1 (Interface Layout), designs should prioritize orderly and harmonious spatial arrangements while avoiding cluttered and conflicting elements to enhance visual comfort and operational efficiency.

Informed by Factor F_2 (Information Presentation), interfaces must ensure prominent and complete information delivery, eliminating hidden or incomplete content to optimize communication efficiency and accuracy. Anchored in Factor F_3 (User Preference), platforms should align with users’ pursuit of novelty and likable elements, while avoiding conventional or dislikable design strategies to maximize satisfaction and loyalty.

Factor analysis indicates that interface layouts require simplified structures and well-defined functional zoning to enhance logical coherence. Designers

should unify visual elements (e.g., color schemes, typography, icons) to avoid clutter. A minimalist style with rationally organized functional areas is recommended to improve readability and accessibility. For instance, optimize menu hierarchies with clear icons and labels.

Information presentation should prioritize rapid user access to critical content. Highlight essential information through high-contrast colors and dynamic visual cues while ensuring completeness and accuracy. Examples include using bold headings and strategic annotations, but designers must balance prominence with layout cleanliness to prevent visual chaos.

To align with user preferences, conduct regular user testing (e.g., A/B testing) to gather feedback and iterate designs. Implement personalized themes and customization options (e.g., dark mode, layout templates) to accommodate diverse aesthetic needs. For example, allow users to switch between “Minimalist Mode” and “Immersive Mode” based on usage contexts.

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