

A Systematic Literature Review on Personalized User Interfaces in Modern Applications

Mukul Kulkarni¹, Om A. Lakade², Tejas S. Kumbhar³, Saad A. Sayyad⁴, Nupur R. Shankhapal⁵

^{1,2,3,4,5}Department of MCA, MES' IMCC, Pune

¹mlk.imcc@mespune.in, ²omlakade8@gmail.com, ³tejasskumbhar2002@gmail.com, ⁴saadsayyed480@gmail.com,

⁵nupurshankhapal@gmail.com

Peer Review Information	Abstract
Type: Article Received: 20 March 2026 Revised: 03 April 2026 Accepted: 21 May 2026 Published: 03 June 2026	This systematic literature review consolidates insights from five seminal research papers to explore the evolution, challenges, and future directions of Personalized User Interfaces (PUI) in adaptive digital ecosystems. It examines the critical distinction between User Interface (UI) design and User Experience (UX) dynamics, emphasizing their interdependence in crafting context-aware interactions. The analysis highlights the paradigm shift from traditional content personalization—exemplified by platforms like Netflix and Spotify—to holistic UI adaptation, incorporating layout, navigation, and interaction patterns driven by machine learning. Key advancements include clustering algorithms (e.g., k-means) for behavioural segmentation, real-world user modelling, and metrics demonstrating enhanced usability (e.g., reduced task completion time) and satisfaction (e.g., improved retention rates). Identified gaps include insufficient integration of personalization principles in design education and limited real-time adaptability to contextual factors like environmental cues. The review proposes AI-driven frameworks for predictive user modelling, emotion-aware interfaces leveraging affective computing, and standardized protocols for balancing user autonomy with automation. These findings underscore the necessity of cross-disciplinary collaboration to address technical and ethical challenges in adaptive design while advancing user-centric digital experiences.
	Keywords: Personalized User Interfaces; Adaptive User Interfaces; User Experience; Machine Learning; Behavioural Segmentation; Human-Computer Interaction; Affective Computing; Context-Aware Systems.

How to Cite This Article

Kulkarni, M., Lakade, O. A., Kumbhar, T. S., Sayyad, S. A., & Shankhapal, N. R. (2026). A systematic literature review on personalized user interfaces in modern applications. *Multidisciplinary Journal of Research in Engineering and Technology*, 13(2), 280-283.

Introduction

As our dependency on digital platforms grows, so does the expectation for these platforms to understand and adapt to individual users. Personalized User Interfaces (PUIs) represent a major shift from standardized designs to systems that mold themselves to a user's behaviour, preferences, and interaction style. While personalization is not new, the scope has expanded—from recommending content to reshaping how interfaces behave and feel. This review brings together findings from multiple domains—education, machine learning, HCI, and mobile app design—to present a comprehensive view of personalization's potential in UI/UX. The research spans from foundational design principles to advanced techniques using clustering algorithms and behavioural modelling.

Methodology

The research involved a thematic synthesis of five academic papers published between 2016 and 2021. The selection was made based on relevance to UI/UX personalization, coverage of practical frameworks, and use of adaptive systems. The papers were sourced using keywords like 'adaptive UI', 'UX personalization', 'machine learning in interfaces', and 'news app customization'. Each paper contributed a unique perspective:

- A foundational study on UI/UX tools and design trends.
- A statistical study analysing student understanding of UI/UX concepts.
- A conceptual model on news app personalization, introducing the 'what vs. how' duality.
- A machine learning-based approach using K-means clustering to dynamically adapt interfaces.
- A user modelling framework proposing reader segmentation (Trackers, Dippers, Reviewers) in mobile applications.

Themes were extracted and categorized under design principles, user behaviour modelling, personalization technologies, and implementation challenges.

Findings & Thematic Analysis

The reviewed papers collectively highlight four major themes: the evolution of UI/UX roles, the importance of adaptive systems, machine learning in personalization, and user-type-based interface design.

UI vs UX and the Importance of Design Thinking

User Interface (UI) encompasses the visual elements—layout, typography, icons—while User Experience (UX) addresses emotions, flow, and efficiency. The first two studies emphasize this distinction and note the lack of awareness in academia, with over 80% of students lacking clarity on implementation. Design thinking and user-centered development must be emphasized in curricula, not just theoretically, but through hands-on, empathy-driven projects.

Tools and Trends Empowering Designers

Figma, Adobe XD, and Sketch are transforming design workflows through collaboration, prototyping, and responsive components. Trends like micro-interactions, minimalism, and gesture-based design dominate the current UX space. However, despite their advancement, these tools still rely on designer intuition and lack built-in personalization intelligence. Integrating real-time user data into these platforms could bridge the gap between static design and adaptive execution.

User Interface Personalization in News Apps

One paper introduces a dual-layer personalization concept—'what' content is shown and 'how' it's accessed. The authors categorize users into Trackers, Reviewers, and Dippers, proposing UI variants for each. For example, a Tracker prefers quick-glance headlines, while Reviewers prefer deep-reading layouts. This model underscores that personalization extends beyond content—it involves aligning interaction design with user intent.

Dynamic Interfaces Using Machine Learning

A study from SRM Institute proposes a machine learning model using K-means clustering to adapt interfaces based on demographics, interaction patterns, and UI design preferences. Their dataset categorized users by age, gender, text weight, and design style (e.g., Google's Material Design vs. iOS styling). The algorithm generated clustered user profiles and UI variants, creating a prototype of three dynamic interfaces. Survey data confirmed user desire for auto-generated personalization rather than manual settings. The study indicates strong user approval for adaptive interfaces that modify themselves over time based on user behavior.

Discussion

The studies collectively argue that personalization must go beyond aesthetics and dive deep into understanding user habits, goals, and even context of usage. While content personalization is widely practiced, UI personalization is still emerging. Adaptive systems have been shown to improve usability, reduce cognitive load, and increase engagement. However, several persistent challenges were identified:

- Static design tools: Most UI/UX tools lack the infrastructure to support real-time, data-driven interface adjustments.
- Educational gaps: Design education often emphasizes visual styling over behavioural personalization, leaving students underprepared for user-adaptive development.
- Surface-level customization: Many systems stop at shallow personalization-like themes or font sizes, neglecting deeper behavioural insights.
- Contextual unawareness: Current personalization models rarely factor in dynamic context—such as environment, device usage conditions, or emotional state.
- Ethical and privacy issues: Collecting and utilizing user data raises concerns around consent, transparency, and control.

The integration of machine learning, especially unsupervised techniques like K-means clustering, is positioned as a scalable solution. These methods allow designers and developers to discover user patterns without requiring labelled datasets, leading to more organic and unbiased personalization strategies. Moreover, the concept of behavioural segmentation—like Trackers, Dippers, and Reviewers—introduces a robust foundation for building adaptive UI structures tailored to different user personas.

Emerging areas like affective computing and emotion-aware interfaces offer a compelling next step. By recognizing mood through biometric or environmental cues (e.g., camerabased emotion detection or ambient sound analysis), future interfaces can adapt not only to *what* a user wants, but *how* they're feeling—enhancing empathy in design.

Conclusion

Personalized User Interfaces are no longer a futuristic vision—they're rapidly becoming essential in today's smart application landscape. From news readers and educational platforms to enterprise tools and e-commerce systems, users now expect digital environments to intuitively respond to their individual needs and preferences.

This review has shown that:

- Design tools are evolving but lack native personalization capabilities.
- User segmentation and machine learning can produce dynamic UIs that adapt in real time.
- Education must evolve to focus on practical, adaptive design thinking.
- Systems should personalize both “what” and “how”—content and interaction—in unison.
- Ethical frameworks are necessary to guide personalization without infringing user privacy or autonomy.

The path forward involves not just refining algorithms or interface styles, but fostering interdisciplinary collaboration—between designers, AI researchers, behavioural scientists, and ethicists. Such cooperation is vital for creating digital experiences that are not only effective but human-cantered, ethical, and inclusive.

Ultimately, the future of user interfaces lies in adaptability: systems that learn from users continuously, respond with empathy, and empower users through intuitive interaction— not overwhelm them with complexity.

References

1. V. Sharma and A. K. Tiwari, "A Study on User Interface and User Experience Designs and its Tools," *World Journal of Research and Review (WJRR)*, vol. 7, no. 5, 2021. Retrieved from https://www.wjrr.org/download_data/WJRR0705020.pdf
2. H. Joo, "Understanding UI and UX and Design Shifts with Interface Changes," *International Journal of Applied Engineering Research*, vol. 12, no. 20, 2017. Retrieved from https://www.ripublication.com/ijaer17/ijaerv12n20_33.pdf
3. M. Constantinides and J. Dowell, "User Interface Personalization in News Apps," *ACM INRA Workshop*, 2016. Retrieved from <https://dl.acm.org/doi/10.1145/2971485.2971501>
4. S. Sankar et al., "Personalized Dynamic User Interfaces," *IOP Conference Series: Materials Science and Engineering*, vol. 1022, no. 1, 2021. Retrieved from <https://iopscience.iop.org/article/10.1088/1757-899X/1022/1/012104>
5. H. Soh et al., "Deep Sequential Recommendation for Personalized Adaptive User Interfaces," *arXiv preprint*, 2017. Retrieved from <https://arxiv.org/abs/1706.07648>
6. G. Adomavicius and A. Tuzhilin, "Toward the Next Generation of Recommender Systems: A Survey of the State-of-the-Art and Possible Extensions," *IEEE Transactions on Knowledge and Data Engineering*, vol. 17, no. 6, pp. 734–749, 2005. Retrieved from <https://ieeexplore.ieee.org/document/1423975>
7. X. Amatriain and J. Basilico, "Netflix Recommendations: Beyond the 5 Stars," *Netflix Tech Blog*, 2015. Retrieved from <https://netflixtechblog.com/netflix-recommendations-beyond-the-5-stars-55838468f429>
8. J. O'Donovan and B. Smyth, "Trust in Recommender Systems," *Proceedings of the 10th International Conference on Intelligent User Interfaces (IUI)*, pp. 167–174, 2005. Retrieved from <https://dl.acm.org/doi/10.1145/1040830.1040841>

9. A. Holzinger, "Interactive Machine Learning for Health Informatics: When Do We Need the Human-in-the-Loop?" *IEEE Intelligent Systems*, 2016. Retrieved from <https://ieeexplore.ieee.org/document/7467408>
10. P. Brusilovsky, "Adaptive Hypermedia," *User Modeling and User-Adapted Interaction*, vol. 11, pp. 87–110, 2001. Retrieved from <https://link.springer.com/article/10.1023/A:1011143116306>
11. F. Ricci, L. Rokach, and B. Shapira, *Recommender Systems Handbook*, 2nd ed. Springer, 2015. Retrieved from <https://link.springer.com/book/10.1007/978-1-4899-7637-6>
12. Y. Koren, R. Bell, and C. Volinsky, "Matrix Factorization Techniques for Recommender Systems," *IEEE Computer*, vol. 42, no. 8, pp. 30–37, 2009. Retrieved from <https://ieeexplore.ieee.org/document/5197422>
13. S. Gauch, M. Speretta, A. Chandramouli, and A. Micarelli, "User Profiles for Personalized Information Access," *The Adaptive Web*, Springer, pp. 54–89, 2007. Retrieved from https://link.springer.com/chapter/10.1007/978-3-540-72079-9_1
14. M. Pazzani and D. Billsus, "Content-Based Recommendation Systems," *The Adaptive Web*, Springer, pp. 325–341, 2007. Retrieved from https://link.springer.com/chapter/10.1007/978-3-540-72079-9_10
15. J. R. Anderson, *Cognitive Psychology and Its Implications*, 7th ed. Worth Publishers, 2010. Retrieved from <https://www.worthpublishers.com>
16. D. Norman, *The Design of Everyday Things*, Revised and Expanded Edition. Basic Books, 2013. Retrieved from <https://www.jnd.org/books/the-design-of-everyday-things-revised-and-expanded-edition.html>
17. J. Nielsen, *Usability Engineering*. Academic Press, 1994. Retrieved from <https://www.nngroup.com/books/usability-engineering/>
18. B. Shneiderman, C. Plaisant, M. Cohen, S. Jacobs, and N. Elmqvist, *Designing the User Interface: Strategies for Effective Human-Computer Interaction*, 6th ed. Pearson, 2016. Retrieved from <https://www.pearson.com/en-us/subject-catalog/p/designing-the-user-interface/P200000003443>
19. ISO 9241-210, *Ergonomics of Human-System Interaction — Part 210: Human-Centred Design for Interactive Systems*. International Organization for Standardization, 2019. Retrieved from <https://www.iso.org/standard/77520.html>
20. A. Cooper, R. Reimann, D. Cronin, and C. Noessel, *About Face: The Essentials of Interaction Design*, 4th ed. Wiley, 2014. Retrieved from <https://www.wiley.com/en-in/About+Face%3A+The+Essentials+of+Interaction+Design%2C+4th+Edition-p-9781118766576>