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Integrating spatio-temporal and psychological perspectives: A new research direction in smart tourism

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Integrating spatio-temporal and psychological perspectives: A new research direction in smart tourism

Tourism is inherently a spatial phenomenon, encompassing a complex assemblage of mobile technologies, infrastructure, online and networked spaces, and the various actors involved in different forms of mobility (Hannam, Butler, & Paris, 2014). Smart tourism refers to tourism that is supported by integrated efforts in an ICT-enabled destination (smart destination), offering ICT-mediated tourism experiences (smart experience), and interconnecting stakeholders within the business ecosystem (smart business) (Gretzel et al., 2015). A significant portion of tourism activities—before, during, and after travel—takes place in electronic markets and on digital/social media platforms, where ICT systems serve as spatial intermediaries between buyers and sellers, as well as among tourists. This is often referred to as 'geospatial smart tourism.' Smart tourism markets can reduce buyer search costs and enable optimal allocation of productive resources (Bakos, 1997). Advances in mobile, social, and location-based technologies have created new opportunities for researchers and practitioners to better manage tourism products and services relevant to space and time, influencing tourists' spatio-temporal movements (Tussyadiah, 2012). Therefore, it is critical to consider space, time, and history when explaining the uneven spatial development of tourism and economic factors, a concept known as spatio-temporal evolutionary economic geography (Jang & Kim, 2022a).

The main research streams in smart tourism focus on several key dimensions, including tourist experience design, psychological perspectives, resident engagement, spatiotemporal analysis, and stakeholder collaboration (see Table 1). First, studies on tourist experience design examine the processes shaping experiences, leveraging geotagged social media and mobile apps for personalization and real-time feedback (e.g., Liu et al., 2022). Second, psychological perspectives are integrated to explore how tourists' traits influence their actual behaviors (e.g., Jang & Kim, 2022b). Third, resident engagement, often underrepresented in research, emphasizes the importance of balancing tourism benefits with community well-being (e.g., Wei et al., 2024). Fourth, spatio-temporal analysis combines conventional statistical data from online booking platforms, and geospatial sources to model and predict tourist behaviors and spatial patterns (e.g., e Silva et al., 2018). Lastly stakeholder collaboration research highlights the need for aligning goals among tourists, firms,

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governments, and residents to achieve sustainable development (e.g., Buhalis & Amaranggana, 2014). However, most studies focus on isolated spatio-temporal or psychological perspectives, providing limited evidence on the causal relationships between psychological factors (e.g., emotions) and spatio-temporal behaviors (e.g., movement patterns). Therefore, smart tourism research demands deeper insights into not only the spatio-temporal ("where + when") patterns of tourism supply, demand, and markets, but also the underlying causes and implications of these phenomena ("why + so what").

| Research stream | Primary focus | Key insights | References | |
|--|---|--|--------------------|--|
| Tourist | Understanding and | GPS and location-based social | Liu et al. (2022); | |
| experience design enhancing tourist journeysnetwork data can be utilized for Tussyad | | | | |
| | and experiences through smart technologies | spatio-temporal destination marketing | | |
| Psychological | Investigating how | Combining geo-psychological and | Jang and Kim | |
| perspectives | psychological traits and | spatio-temporal data offers new | (2022b); Teo | |
| | perspectives influence | research opportunities | (2015) | |
| | tourist decision-making | | | |
| Resident | Examining the role of | Smart tourism destinations should | Wei et al. (2024) | |
| engagement | local residents in shaping smart tourism ecosystems | increase travellers' experiences and sresidents' quality of life | | |
| Spatio-temporal | Understanding tourists' | GPS, geotagged content, and big data | e Silva et al. | |
| analysis | movements and behaviors in space and time | s model spatio-temporal patterns and segment tourist behaviors | (2018) | |
| Stakeholder | Exploring interactions | ICT enables interconnectedness but | Buhalis and | |
| collaboration | between tourists, tourism | requires aligning diverse stakeholder | Amaranggana | |
| | firms, governments, and | goals | (2014); Gretzel et | |
| | other stakeholders | | al. (2015) | |

| Table 1. Main streams of | research in smart | tourism |
|--------------------------|-------------------|---------|
|--------------------------|-------------------|---------|

Critical psychology suggests that individual subjectivity, such as that of tourism firms and tourists, is deeply embedded in social, cultural, and historical contexts (Teo, 2015). This indicates that spatial configurations are intrinsically linked to how tourism firms and tourists perceive, decide, and behave. Geo-psychological research, which explores geographical differences in psychological characteristics, integrates both the causes and consequences of these variations. This type of research often presents challenges, requiring scholars to navigate unfamiliar data formats, sources, measures, and statistical issues unique to geographical analysis (Ebert et al., 2022). To date, no research has addressed geopsychological issues in smart tourism. Therefore, future studies should investigate the causes and impacts of tourists' spatio-temporal behaviors at specific destinations by collecting both behavioral (secondary) and intentional (primary) data and analyzing them through a causal lens.

Methodologically, emerging geospatial data sources and advanced analytical techniques offer significant opportunities for tourism scholars to analyze tourist and tourism activities from a geospatial perspective. Recent advances in ICT and the evolution of geospatial data are reflected in the increasing use of various big data sources that integrate geographic information, such as geotagged social media and mobile phone data. Geographic information system (GIS)-based analytical techniques and spatial econometric methods are also powerful tools for addressing complex spatial decision problems in tourism (Koo et al., 2023). The spatial analysis of tourists' movements and tourism supply and demand has long been a core component of the geographical approach to tourism (Kang, Kim, & Nicholls, 2014), given that tourism industry can arise from both supply and demand sides (Yang & Mao, 2019). The availability of various forms of big data—such as text, images, location data, and video—in tourism provides vast opportunities for researchers and practitioners to explore physical and online tourism markets, tourism supply, and tourists' perceptions and behaviors in a more holistic and sophisticated manner.

Recently, geospatial artificial intelligence (GeoAI), which refers to spatially explicit AI techniques (Janowicz et al., 2020), has emerged as a powerful tool by integrating spatiotemporal data with machine learning and deep learning methods to extract insightful knowledge (VoPham et al., 2018). GeoAI can develop intelligent computer programs that mimic human perception and spatial reasoning processes, allowing for a deeper understanding of geographical phenomena and the ability to solve complex problems involving human-environmental systems and their interactions (Gao, 2021). Consequently, applications of GeoAI can accelerate the real-world understanding of the smart tourism field by providing valuable place-based policy implications, addressing the critical question of 'so what' in the context of spatio-temporal analysis of tourism firms and tourist behaviors.

Given the multifaceted nature of smart tourism, we propose a conceptual framework that integrates spatio-temporal, psychological, and methodological perspectives (Figure 1). This framework builds on critical psychology, which posits that spatial configurations including social, cultural, and historical contexts—inform psychological perspectives and, in turn, shape individual behaviors (Teo, 2015). We extend this geo-psychological approach to contemporary smart tourism systems, where tourist perceptions and behaviors are

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increasingly identifiable through spatio-temporal big data and advanced analytical techniques (Ebert et al., 2022).

The first area of integration, *spatio-temporal analytics*, involves the use of finegrained spatio-temporal data to develop sophisticated models for understanding tourist behaviors. As technology evolves, new data types and advanced analytical methods enable a deeper understanding of these behaviors (Caldeira & Kastenholz, 2020). For example, Li et al. (2019) combined GPS data with survey responses from Chinese tourists to identify factors influencing destination choices, while Liu et al. (2022) analyzed open GPS-trajectory data to examine the spatio-temporal movement patterns of tourists at Mount Huashan, China, clustering them into three distinct tourist segments.

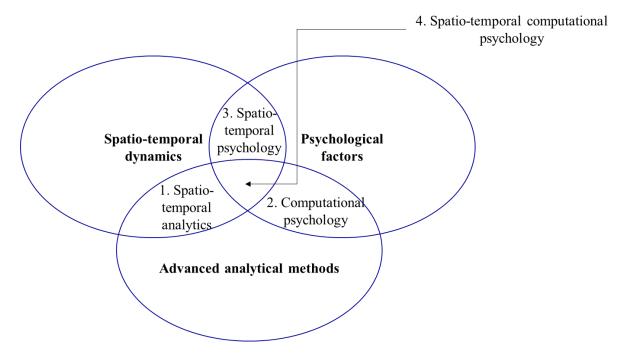


Figure 1. Conceptual framework of spatio-temporal-psychological approach in smart tourism

The second area focuses on *computational psychology* in tourism research, aiming to improve predictive accuracy and understand causal relationships (Hofman, Sharma, & Watts, 2017; Lazer et al., 2009). Researchers apply natural language processing, statistical techniques, and machine learning to analyze sentiment from large textual datasets, such as social media reviews (Cambria et al., 2017). For instance, Xiang et al. (2017) highlighted the challenges of predictive accuracy in social media analytics, incorporating data from platforms like TripAdvisor, Expedia, and Yelp, while Fu et al. (2018) used a meta-learning approach to explore the effects of design on predictive accuracy through sentiment analysis of Chinese travel news.

The third area, known as *spatio-temporal psychology*, merges psychology with spatiotemporal data in tourism studies. Some research, such as that by Jang and Kim (2022b), has explored the causes of spatial tourism behavior by integrating psychological mechanisms. These studies often use experimental methods, similar to those employed by Jang et al. (2021), alongside geospatial data analysis. However, most of this research relies on crosssectional data, which limits the understanding of how psychological factors interact with longitudinal geospatial data. Consequently, there remains a gap in empirical research that combines geo-psychological factors with spatio-temporal dynamics in tourism. Addressing this gap could offer a more comprehensive view of tourist behavior over the long term.

The fourth area, *spatio-temporal computational psychology*, integrates granular spatio-temporal tourist mobility data with psychological information and advanced analytics (Lazer et al., 2009). While earlier studies have attempted to combine GPS data with survey responses to understand destination choices (e.g., Li et al., 2019), incorporating computational psychology techniques offers the poential to enhance the granularity and depth of findings (Fu et al., 2019; Hofman et al., 2017). However, this integration has faced challenges, including limited recognition of geo-spychological interactions, inadequate data integration, and the complexity of analyzing multi-dimensional data. AI techniques, such as deep learning and natural language processing (NLP), provide powerful tools to analyze large-scale data and uncover insights into the interplay between psychological states and spatio-temporal behaviors. Advances in data collection, analytical methods, and interdisciplinary collaboration are steadily overcoming these obstacles, paving the way for more holistic and dynamic models of tourist behavior.

In this research note, we encourage researchers and practitioners to integrate spatiotemporal dynamics, psychological factors, and advanced analytical methods in smart tourism research. First, smart tourism should examine how firms design and implement spatiotemporal products and services, evaluating their impact on performance. These insights can enhance customer satisfaction through personalized experiences, such as recommending lowcrowd activities or emotionally significant locations, while optimizing revenue and fostering loyalty. Second, consumer behavior related to spatio-temporal services should be explored from a psychological perspective. Studying how spatio-temporal-psychological offerings influence attitudes and decision-making will help firms predict tourist reactions, design effective marketing, and build trust in tech-driven solutions. Third, researchers could develop effective tourism policies from a spatio-temporal-psychological perspective, particularly in the context of regional or global crises (e.g., earthquakes and pandemics). Such policies could

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help mitigate the negative impacts on tourism and aid in recovery by considering how tourists and tourism providers psychologically respond to spatio-temporal dimensions during these events.

Finally, advancing research methodologies in smart tourism is crucial. Future studies could focus on creating tools to measure spatio-temporal-psychological tourist engagement, combining primary (intentional) and secondary (behavioral) data for a more comprehensive understanding of these dynamics. Advanced modeling techniques could also be employed to analyze spatio-temporal-psychological tourist behaviors, with efforts to improve the interpretability of GeoAI models in these contexts. Refining these methodologies would offer more actionable insights and better support the development of intelligent, adaptive, and resilient smart tourism systems.

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