

1 Artificial Intelligence in the Museum Experience: 2 Comparative Perspectives from Beijing, Turin, and 3 Harvard

5 This article examines how artificial intelligence is transforming the
6 contemporary museum from a static repository of objects into a dynamic
7 system of interpretation, participation, and imagination. Through a
8 qualitative comparative analysis of three emblematic case studies from
9 Beijing, Turin, and Cambridge, the study explores how different cultural,
10 political, and epistemological frameworks shape the integration of AI in
11 museological practice. The Beijing case illustrates a state-led infrastructural
12 model in which AI supports large-scale heritage governance and digital
13 sovereignty. The Turin case highlights a participatory and human-centered
14 approach, where AI functions as a mediator aligned with ethical design,
15 community engagement, and sustainability. The Cambridge case, represented
16 by Harvard University's Chinese Art Media Lab, presents an experimental
17 paradigm in which AI operates as a creative and imaginative partner in
18 immersive reconstruction and algorithmic aesthetics. Drawing on digital
19 hermeneutics, phenomenology, and actor-network theory, the article argues
20 that AI acts simultaneously as technological infrastructure and interpretive
21 agent, redistributing authority among curators, visitors, and machines. The
22 intelligent museum thus emerges as a cognitive ecosystem where cultural
23 meaning is co-produced through human-machine collaboration, raising new
24 ethical, epistemological, and aesthetic questions for the future of digital
25 museology.

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27 **Keywords:** Artificial Intelligence; Digital Museology; Cultural Heritage;
28 Human Computer Interaction; Immersive Media

30 31 **Introduction**

32
33 The twenty-first-century digital transformation of museums can be used to
34 showcase how societies are reshaping the creation and dissemination of cultural
35 knowledge using computational technologies (Nanetti, Razdi, & Benvenuti,
36 2021). Artificial intelligence has become a vital element in this process,
37 widening the museum's role from merely storing objects to actively engaging in
38 content interpretation and co-creation of new knowledge. The incorporation of
39 intelligent systems into museum practice alters both institutional logic and
40 visitor experience, introducing new forms of curation, conservation, and
41 narrative interaction (Parry & Dziekan, 2021).

42 Museums have historically mediated between memory and imagination,
43 between the tangible presence of artifacts and the symbolic frameworks that give
44 them meaning (Hooper Greenhill, 1992; Nanetti, 2021; Murawska-Muthesius &
45 Piotrowski, 2015). With the rise of artificial intelligence, the museum becomes
46 a dynamic network where objects, data, and human experiences interact
47 (Caramiaux, 2023). This transformation reflects what Fiona Cameron describes

1 as a broader heritage complex, in which digital objects and technologies are
2 embedded within an institutionalized culture of practices and ideas that actively
3 shape value, interpretation, and mediation in museums (Cameron, 2007, p. 51).
4 In 2021, Andrea Nanetti, Zaheer Razdi, and Davide Benvenuti reviewed the
5 secondary literature in English on the strengths and weaknesses of web-based
6 learning tools in museum systems, as highlighted by the COVID-19 pandemic
7 in 2020 (Nanetti, Razdi, & Benvenuti, 2021).

8 Artificial intelligence contributes to this evolution by enabling predictive
9 conservation, automated classification, and adaptive storytelling that adjusts to
10 audience behavior (Caramiaux, 2023). The redefinition of the museum through
11 artificial intelligence signifies not only an institutional change but also a
12 philosophical one. The lines between material and digital heritage are blurring
13 as algorithms contribute to the creation of meaning. As Rafael Capurro explains,
14 digital culture necessitates a hermeneutic approach where understanding
15 develops through the interaction between human and artificial intelligences
16 (Capurro, 2010). Museums can therefore be seen as dynamic cognitive
17 ecosystems in which visitors, curators, and intelligent systems collaboratively
18 generate experiences and knowledge.

19 Our research team compared three notable case studies from different
20 cultural contexts and museographical approaches (Beijing, China; Turin, Italy;
21 and Cambridge, Mass., USA) to demonstrate how artificial intelligence is
22 reshaping the technological, ethical, and epistemological roles of cultural
23 institutions. Each case study embodies a distinct paradigm of technological
24 integration. The Beijing case study illustrates a state-led model in which artificial
25 intelligence serves as cultural infrastructure, integrating heritage preservation
26 with national digital strategies (Peng, 2022). The Turin case study examines the
27 I-Muse project, which embodies a participatory, human-centered model that
28 links artificial intelligence to ethical design, community engagement, and
29 environmental sustainability (Politecnico di Torino, 2023). The Cambridge case
30 study highlights Harvard University's Chinese Art Media Lab (CAMLab),
31 which offers an experimental aesthetic model in which artificial intelligence
32 serves as a creative partner in reconstructing lost cultural experiences through
33 immersive visualization (Kenderdine, 2021).

34 The comparison of these three models highlights the diversity of approaches
35 to artificial intelligence in museology and the different cultural, political, and
36 ethical assumptions underlying them. It addresses three central research
37 questions. First, how does artificial intelligence reshape the epistemological and
38 ethical dimensions of museums? Second, how do different sociocultural
39 frameworks influence the adoption and interpretation of intelligent technologies?
40 Third, what theoretical paradigms emerge from the comparative study of Eastern,
41 European, and North American experiences? By situating artificial intelligence
42 within the broader field of cultural production, this article aims to develop a
43 critical theory of the intelligent museum. The analysis is rooted in the
44 intersection of media studies, cultural theory, and digital humanities. It argues
45 that integrating artificial intelligence into museology is not merely a technical
46 process but a redefinition of the museum's communicative and cognitive

1 functions. As Elisa Giaccardi notes, in a participatory culture heritage is less a
 2 static collection of artifacts than an ongoing, socially constructed process
 3 sustained through repeated interactions among people, places and digital media
 4 (Giaccardi, 2012, pp. 1–3).

5 In this context, artificial intelligence serves as both a medium and a
 6 metaphor. It enables museums to extend their reach beyond physical boundaries
 7 and to engage audiences in new ways, while also reflecting broader
 8 transformations in how societies conceptualize knowledge, memory, and
 9 creativity. The comparative perspective adopted here seeks to show that the
 10 museum is not only adapting to artificial intelligence but also contributing to its
 11 cultural meaning. Through this dialogue between technology and interpretation,
 12 the museum becomes a site where human understanding and machine cognition
 13 meet in the continuous re-creation of heritage.

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16 **Methodology and Theoretical Framework**

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18 The methodological structure of this study is based on a qualitative
 19 comparative approach that integrates interpretive analysis, critical hermeneutics,
 20 and actor-network theory. This study examines how artificial intelligence
 21 interacts with institutional, technological, and cultural systems across different
 22 national contexts. Instead of measuring the quantitative impact of technology,
 23 this research investigates how meaning, participation, and authority are
 24 renegotiated through intelligent mediation within museums (Latour, 2005).

25 Each of our three case studies (Beijing, Turin, and Cambridge) represents a
 26 distinct configuration of technology, governance, and cultural ideology. The
 27 study adopts a comparative case analysis method that enables the identification
 28 of convergences and divergences among these systems, showing how the
 29 museum's epistemological role changes under artificial intelligence. The
 30 approach is interpretive rather than positivist; it seeks to understand how
 31 artificial intelligence and automated systems operate as agents with their own
 32 forms of agency, intelligence, and cognition, reshaping the relations between
 33 humans and heritage (Cameron, 2021, pp. 3-8).

34 The first theoretical axis grounding this analysis is digital hermeneutics,
 35 which views understanding as an interactive process between human cognition
 36 and technological systems (Nanetti, 2023, 30-60). In this framework, artificial
 37 intelligence is not a neutral instrument but an interpretive partner that generates
 38 and transforms cultural meaning. Rafael Capurro argues that digital
 39 hermeneutics requires recognizing the co-agency of technology in the
 40 construction of knowledge, since interpretation increasingly occurs through
 41 algorithmic mediation (Capurro, 2010). Applied to museology, this means that
 42 algorithms, databases, and recommendation systems participate in the creation
 43 of historical and aesthetic narratives.

44 The second axis is actor-network theory, which conceptualizes museums as
 45 networks composed of human and nonhuman actors. Curators, visitors, digital
 46 interfaces, sensors, and machine learning models all contribute to the

1 construction of museum discourse. Artificial intelligence, within this perspective,
2 is not external to the museum but part of its communicative and ontological
3 fabric. Latour defines this relational ontology as one in which agency circulates
4 among interconnected entities, dissolving the boundaries between subject and
5 object (Latour, 2005). The museum thus becomes a distributed field of
6 interpretation where knowledge emerges from interaction rather than instruction.

7 The third theoretical axis is phenomenology of experience, emphasizing that
8 perception is the foundation of understanding. The French phenomenological
9 philosopher Maurice Merleau-Ponty (1908-1961), in his seminal book
10 *Phénoménologie de la perception* (1945), situates meaning in the embodied
11 encounter between the subject and the world (Merleau-Ponty, 2010). Artificial
12 intelligence expands this phenomenological domain by introducing interfaces
13 that sense, respond, and adapt to human presence. In contemporary museums,
14 interactive and socially networked media can shift visitors from passive
15 spectators to active co-creators of meaning and shared experience (Giaccardi,
16 2012). Through these interactions, visitors no longer passively receive
17 information but can participate in the production of cultural significance.

18 Our research combined documentary analysis, literature review, and case
19 observation. Primary materials include institutional reports, exhibition records,
20 and policy documents from the museums under examination. Secondary
21 materials include recent peer-reviewed literature in museum studies, digital
22 heritage, and artificial intelligence ethics (Peng, 2022; Zhu & Liu, 2025). These
23 sources provide empirical and theoretical grounding for understanding how
24 intelligent systems reshape conservation, accessibility, and participation in
25 cultural institutions. The integration of these methodologies allows the research
26 to go beyond merely descriptive analysis. By examining technological
27 infrastructures alongside philosophical and social dimensions, the study situates
28 artificial intelligence within a broader cultural framework. This interpretive
29 approach recognizes that technology is never purely instrumental but always
30 embedded in systems of meaning. Artificial intelligence in museums, therefore,
31 functions as both a practical tool and a theoretical challenge, questioning
32 traditional notions of authorship, authenticity, and authority.

33 Finally, our methodology approach follows a hermeneutic cycle, moving
34 between the particular and the general. Insights from each case study inform the
35 overall theoretical model, which in turn reframes the interpretation of local
36 practices. This cyclical process reflects the mutual shaping of theory and
37 observation. It also aligns with the epistemological assumption that artificial
38 intelligence, as both an analytical and creative agent, modifies not only how
39 museums operate but how they think. The museum, in this framework, is
40 conceptualized as a living system where human interpretation and algorithmic
41 reasoning interact continuously in the co-production of knowledge.

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1 **Beijing: Artificial Intelligence as Infrastructure**

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3 In Beijing, the application of artificial intelligence to the museum sector
 4 reflects a distinctive model in which technological innovation and cultural policy
 5 are integrated within a unified state vision. Artificial intelligence has become a
 6 structural component of China's cultural modernization, linking the preservation
 7 of national heritage to the strategic goal of digital sovereignty. Museums in
 8 Beijing thus operate as laboratories for testing the role of intelligent technologies
 9 in governance, education, and the formation of social identity (Peng, 2022).

10 The Chinese government has identified cultural heritage as a key domain
 11 for the implementation of artificial intelligence. National programs such as
 12 “Digital China” and the “Smart Museum Initiative” promote the use of machine
 13 learning, computer vision, and data analytics to manage vast collections and
 14 enhance public access. The Palace Museum, the National Museum of China, and
 15 the Capital Museum have all experimented with AI-driven tools such as image
 16 recognition, 3D modelling and advanced search interfaces to support restoration,
 17 cataloguing and interpretation of large collections, in line with broader
 18 developments in “smart museum” services in China (Peng, 2022). These
 19 initiatives transform the museum into a complex digital infrastructure where
 20 algorithms govern not only the classification of artifacts but also the ways in
 21 which audiences encounter them.

22 Artificial intelligence aids preventive conservation by constantly
 23 monitoring environmental conditions and detecting micro-deterioration. In
 24 collaboration with the Chinese Academy of Cultural Heritage, several Beijing
 25 museums have implemented intelligent sensors that gather real-time data on
 26 humidity, temperature, and light exposure. The integration of these systems has
 27 lowered restoration costs and improved the accuracy of conservation planning
 28 (Zhu & Liu, 2025). Through these mechanisms, artificial intelligence enhances
 29 the museum's ability to preserve material culture while also creating extensive
 30 data archives that contribute to national digital resources. At the same time, the
 31 Beijing model demonstrates how artificial intelligence fits within a broader
 32 ideological framework. The automation of curation and the personalization of
 33 digital experiences are aimed at strengthening the narrative of cultural continuity
 34 and national unity. Algorithms not only recommend exhibitions or generate
 35 multilingual captions but also shape interpretive hierarchies that align with state-
 36 approved historical narratives (Peng, 2022). Artificial intelligence thus functions
 37 both as a mechanism of access and as an instrument of cultural governance.

38 The Palace Museum's intelligent guide services exemplify this duality. In
 39 line with wider developments in smart museums, experimental systems combine
 40 computer vision, facial recognition, natural language processing and behavioural
 41 analytics to tailor information and routes to different visitor profiles, while
 42 simultaneously collecting data that can be used to redesign circulation patterns
 43 and exhibition layouts. This combination of personalization and surveillance
 44 reveals a tension inherent in Beijing's technological paradigm: the museum is
 45 both an open educational space and a managed environment of information
 46 exchange (Wen & Ma B, 2024).

1 From a theoretical perspective, the Beijing model demonstrates how
 2 artificial intelligence transforms the museum into a cybernetic institution. It
 3 aligns with Latour's concept of the networked actor, in which agency is
 4 distributed among humans, machines, and institutions (Latour, 2005). The
 5 museum's curatorial authority becomes a system of coordination among
 6 technical and human actors rather than an expression of individual expertise.
 7 Curators rely on predictive algorithms to identify audience interests, while
 8 administrators use analytics dashboards to allocate resources and measure
 9 engagement. In this environment, artificial intelligence becomes a co-author of
 10 cultural interpretation. This infrastructural paradigm has significant implications
 11 for accessibility and international collaboration. Online exhibitions powered by
 12 artificial intelligence, such as the "Virtual Forbidden City," have attracted
 13 millions of visitors worldwide, demonstrating the capacity of Chinese museums
 14 to expand cultural influence through digital diplomacy (Zhu & Liu, 2025). The
 15 project's immersive environments and AI-driven translation systems enable
 16 cross-cultural interaction, while simultaneously projecting a curated image of
 17 national heritage consistent with policy objectives.

18 The strength of the Beijing model lies in its coherence and scale. It mobilizes
 19 public institutions, universities, and technology companies in a shared effort to
 20 digitize heritage and educate citizens through intelligent media. Its limitation
 21 resides in the restricted openness of interpretation, which can reduce the
 22 diversity of perspectives and the autonomy of audiences. Artificial intelligence,
 23 in this context, reinforces the institutional framework rather than destabilizing it.
 24 Nevertheless, the model provides a powerful example of how technology can
 25 serve both cultural preservation and state strategy. The Beijing case reveals that
 26 artificial intelligence operates as infrastructure, governance, and ideology
 27 simultaneously. It integrates technological efficiency with symbolic production,
 28 turning the museum into an interface between culture and policy. Understanding
 29 this system is essential for evaluating the global future of artificial intelligence
 30 in museology, as it highlights both the possibilities of innovation and the risks of
 31 centralization. In Beijing, the intelligent museum is not only a site of knowledge
 32 but a site of power, where data and heritage converge in the construction of a
 33 digital civilization (De Masi, 2025).

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36 **Turin: Artificial Intelligence as Mediation**

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38 In contrast to Beijing's centralized and policy-driven approach, the Turin
 39 model represents a participatory and decentralized vision of how artificial
 40 intelligence can transform the museum experience. Italian museology has
 41 historically emphasized cultural heritage as a living process that connects
 42 communities, education, and place. In this participatory framework, digital
 43 technologies are conceived less as instruments of control and more as mediators
 44 of meaning within collaborative, human-centred heritage practices (Giaccardi,
 45 2012). The most emblematic example of this approach is the I-Muse project,
 46 developed by the *Politecnico di Torino* in collaboration with several regional

1 museums and cultural organizations. The initiative integrates artificial
2 intelligence and Internet of Things technologies to enhance accessibility,
3 environmental sustainability, and audience participation (Politecnico di Torino,
4 2023). Through adaptive algorithms, the system analyzes visitor behavior and
5 adjusts narrative content, lighting, and temperature to create a more responsive
6 environment. The approach does not seek to predict or govern human behavior
7 but to establish a dialogical relationship between visitors and space.

8 The Italian framework aligns closely with the European Union's Ethics
9 Guidelines for Trustworthy Artificial Intelligence, which stress transparency,
10 accountability, and human oversight (European Commission, 2021). These
11 principles shape both the technological architecture and the philosophical
12 foundations of projects like I-Muse. Artificial intelligence is treated as a
13 collaborator in interpretation, assisting curators and educators in organizing
14 content while preserving the human dimension of decision-making. This
15 approach contrasts sharply with models that prioritize efficiency over
16 participation, positioning ethical design as an integral component of digital
17 transformation.

18 Participation in Turin extends beyond the museum's physical boundaries.
19 Local schools, universities, and cultural associations contribute to the creation
20 of digital materials, ensuring that communities remain active producers of
21 heritage. Workshops and training programs encourage citizens to generate new
22 content and reinterpret existing collections, making the museum a participatory
23 hub of cultural production (Russo, , 2025). This co-creative structure supports
24 what Ross Parry defines as the "distributed museum," a networked ecosystem
25 where authority is shared among institutions, professionals, and the public (Parry,
26 2010; Parry & Dziekan, 2021). Artificial intelligence also contributes to
27 environmental and operational sustainability. Machine learning algorithms
28 monitor energy consumption, predict maintenance needs, and optimize climate
29 control systems, reducing both costs and ecological impact (Politecnico di
30 Torino, 2023). In this sense, technological innovation becomes part of a broader
31 ecological ethic that connects digital efficiency to environmental awareness. The
32 intelligent museum in Turin thus emerges as a hybrid system in which cultural,
33 social, and ecological sustainability reinforce each other.

34 A key characteristic of the Turin model is its interdisciplinary collaboration.
35 Engineers, designers, historians, and sociologists work together throughout the
36 design process, ensuring that artificial intelligence applications reflect both
37 technical feasibility and cultural sensitivity. The design process is iterative and
38 participatory, incorporating user and stakeholder feedback, so that digital
39 technologies support an ongoing, reflexive relationship in which heritage
40 institutions learn from their interactions with the public (Giaccardi, 2012).
41 Artificial intelligence becomes not only a tool for personalization but also a
42 mirror that reveals how institutions engage with their communities.

43 From a theoretical perspective, the Turin model embodies what Rafael
44 Capurro describes as the hermeneutic dimension of digital culture, in which
45 understanding emerges from continuous interpretation between humans and
46 machines (Capurro, 2010). Algorithms in I-Muse do not replace curatorship but

1 extend it, providing insights into audience engagement that curators reinterpret
 2 within broader cultural narratives. This process preserves the interpretive
 3 autonomy of human actors while recognizing the analytical power of intelligent
 4 systems. The result is a balanced model of co-agency between technology and
 5 human creativity. Ethical responsibility remains central to this model.
 6 Transparent data management, informed consent, and algorithmic explainability
 7 are treated as non-negotiable conditions for technological adoption. These
 8 practices reflect a European commitment to cultural democracy and digital
 9 citizenship (European Commission, 2021). In Turin, artificial intelligence is not
 10 merely a technical innovation but an ethical statement about the values that guide
 11 the relationship between culture and technology.

12 The Turin case demonstrates that artificial intelligence can enrich rather than
 13 replace human mediation. Its focus on participation, transparency, and
 14 sustainability positions it as a counterpoint to Beijing's infrastructural model. If
 15 the Chinese museum represents artificial intelligence as a system of governance,
 16 the Italian museum represents it as a space of dialogue. In both cases, technology
 17 transforms the relationship between culture and public, but the Turin model
 18 shows that such transformation can remain deeply human, inclusive, and
 19 ethically grounded.

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22 Cambridge: Artificial Intelligence as Imagination

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24 The case of Cambridge, represented by the Chinese Art Media Lab
 25 (CAMLab) at Harvard University, illustrates a third paradigm in the relationship
 26 between artificial intelligence and the museum: the paradigm of imagination.
 27 Unlike the infrastructural orientation of Beijing or the participatory mediation of
 28 Turin, CAMLab approaches artificial intelligence as a philosophical and
 29 aesthetic instrument. Here, technology is not limited to management or
 30 accessibility but becomes a creative medium for reinterpreting historical
 31 consciousness and visual culture (Kenderdine, 2021).

32 CAMLab's research explores how digital reconstruction and algorithmic
 33 visualization can reanimate lost or fragmented cultural experiences. Projects
 34 such as the Cave Dance project and the Embodied Architecture reconstruction
 35 of the Shakyamuni Pagoda harness artificial intelligence to synthesize data from
 36 mural depictions, archaeological surveys, photographic archives, and art
 37 historical documentation. These reconstructions use machine-learning models
 38 trained on datasets of Dunhuang mural figures and motion-capture recordings of
 39 professional dancers to generate human-computer collaborative choreographies
 40 and immersive environments that allow audiences to experience ancient artistic
 41 spaces in new ways. The result is not a simple replication of historical reality but
 42 a creative translation of it into new sensory and cognitive forms (Cavedance,
 43 2021).

44 In this framework, artificial intelligence acts as a co-creator. Algorithms
 45 analyze patterns of color, motion, and composition to generate plausible
 46 reconstructions that extend the interpretive process beyond human limitation.

1 This collaboration between human imagination and computational inference
 2 exemplifies what Kenderdine defines as “algorithmic aesthetics,” a mode of
 3 artistic production in which data becomes a medium of expression (Kenderdine,
 4 2021). The museum is increasingly conceptualized as a site of simulation and
 5 performance, where visitors engage with cultural heritage not as passive
 6 spectators but as participants in a process of re-creation.

7 CAMLab’s work is grounded in the philosophy of digital phenomenology,
 8 which treats technological mediation as an extension of perception. The
 9 immersive installations produced by the laboratory do not aim to reproduce the
 10 past but to evoke its experiential essence. Through artificial intelligence, the act
 11 of viewing becomes an act of remembering and imagining at once. As Bolter and
 12 Grusin argue, digital media function through remediation, continually translating
 13 old forms into new interfaces (Bolter & Grusin, 1999). CAMLab’s installations
 14 exemplify this principle by transforming the archive into an event of presence,
 15 where history becomes performative.

16 The laboratory’s interdisciplinary orientation reinforces its theoretical depth.
 17 Collaborations among art historians, computer scientists, engineers, and
 18 philosophers foster a research environment that bridges humanistic inquiry and
 19 technical experimentation. This synthesis embodies what Ross Parry calls the
 20 “epistemology of participation,” in which knowledge emerges through the
 21 interaction between disciplines, tools, and sensibilities (Parry, 2010; Parry &
 22 Dziekan, 2021). Artificial intelligence, within this ecology, operates as a
 23 cognitive partner that amplifies rather than replaces human creativity.

24 One of CAMLab’s most significant contributions lies in its exploration of
 25 AI-driven storytelling. Using natural-language generation and visual synthesis,
 26 the lab creates adaptive narratives that evolve in real time based on audience
 27 input. These interactive systems can be understood as what Cameron and
 28 Kenderdine call “multi-perspectival narratives”, in which meaning is not
 29 predetermined but dynamically constructed (Cameron & Kenderdine, 2007).
 30 The museum thus becomes a dialogical field of interpretation, where visitors
 31 negotiate their own understanding through continuous engagement with data and
 32 simulation.

33 The imaginative potential of this model raises critical philosophical
 34 questions about authenticity and authorship. Traditional museology associates
 35 authenticity with the originality of objects and the authority of curators. In
 36 contrast, the AI-driven museum redefines authenticity as a relational quality that
 37 emerges through interaction. The German philosopher and media theorist Walter
 38 Benjamin (1892-1940), reflecting on the “aura” of the artwork, suggested that
 39 technological reproduction diminishes the uniqueness of artworks (Benjamin,
 40 2019). Within the context of artificial intelligence, this same aura can instead be
 41 reinterpreted as the affective intensity generated by the digital presence.
 42 Authenticity, in this sense, resides not in the object itself but in the network of
 43 relations that constitutes its meaning.

44 CAMLab’s projects also engage with ethical and epistemological issues
 45 surrounding the simulation of cultural heritage. The use of predictive algorithms
 46 to fill gaps in incomplete artifacts raises questions about interpretation and

1 responsibility that resonate with broader debates on AI-driven restoration and
 2 neural rendering in cultural heritage. Rather than treating algorithmic outputs as
 3 neutral reconstructions, CAMLab foregrounds the speculative nature of its
 4 interventions, emphasizing that any digital restoration or immersive
 5 visualization is a reversible, virtual layer added to the historical record. The lab
 6 addresses these challenges by maintaining transparency about its methods,
 7 documenting datasets, workflows, and aesthetic choices, and explicitly
 8 presenting its installations as interpretive experiments rather than definitive
 9 restitutions. This reflexivity positions CAMLab's practice within the emerging
 10 field of critical digital heritage, which insists on linking technical innovation
 11 with accountability, explainability, and respect for the plurality of historical
 12 interpretations (Colace, 2025).

13 The Cambridge model demonstrates that artificial intelligence can expand
 14 the museum's ontological boundaries. It shifts attention from the preservation of
 15 material objects to the cultivation of experiential knowledge. The integration of
 16 immersive environments, data visualization, and interactive storytelling
 17 transforms the museum into a space of epistemic experimentation. This aligns
 18 with Rafael Capurro's concept of hermeneutic dialogue between human and
 19 artificial cognition, where meaning arises from reciprocal interpretation
 20 (Capurro, 2010). The museum becomes an arena, in which the past, the present,
 21 and the virtual coexist within a single dynamic field of imagination.

22 In contrast to Beijing's emphasis on infrastructure and Turin's focus on
 23 mediation, Cambridge proposes an aesthetic and speculative engagement with
 24 artificial intelligence. Its strength lies in the ability to transform data into
 25 experience and experience into thought. By integrating art, science, and
 26 philosophy, CAMLab articulates a new role for the museum in the digital age:
 27 not as a container of memory but as a generator of possible worlds. The
 28 laboratory's practice suggests that the future of museology may depend not only
 29 on how technology preserves the past but on how it enables humanity to
 30 reimagine it.

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33 Comparative Insights and Global Implications

34

35 The comparative analysis of Beijing, Turin, and Cambridge reveals that
 36 artificial intelligence functions simultaneously as infrastructure, mediation, and
 37 imagination. These three models correspond to different configurations of
 38 cultural values and institutional priorities; however, they share a common
 39 tendency to redefine the museum as a dynamic network of human and nonhuman
 40 actors. The convergence of algorithmic systems, sensory environments, and
 41 participatory design demonstrates that artificial intelligence is not merely a
 42 technological upgrade but reshapes the epistemological foundations of museums
 43 and digital heritage (Cameron, 2021).

44 In Beijing, artificial intelligence operates as a technological and political
 45 infrastructure. It embodies the logic of centralized coordination, aligning cultural
 46 preservation with national strategy. The emphasis is on scale, efficiency, and the

1 integration of heritage into the broader framework of digital governance.
 2 Algorithms manage vast datasets and construct narrative coherence within a
 3 state-defined horizon. This system reinforces cultural identity by synchronizing
 4 narratives of heritage with narratives of modernization, producing a museum that
 5 exemplifies technological mastery yet remains constrained by institutional
 6 hierarchy.

7 In Turin, artificial intelligence serves as a mediator. It connects technology
 8 with ethics, data with community, and innovation with sustainability. The I-Muse
 9 project demonstrates that artificial intelligence can operate as a dialogical tool
 10 that enhances transparency and inclusion rather than authority (Politecnico di
 11 Torino, 2023). This model embodies a European humanistic tradition that
 12 prioritizes participation, empathy, and ecological responsibility. It reflects the
 13 belief that technology should enrich human experience rather than replace it.

14 Cambridge, by contrast, represents artificial intelligence as imagination.
 15 Through CAMLab's research, the museum becomes a site of creative
 16 speculation and philosophical inquiry. Algorithms reconstruct lost cultural
 17 experiences, generating immersive encounters that blend historical fidelity with
 18 artistic experimentation (Kenderdine, 2021). The Cambridge model thus extends
 19 museology beyond preservation toward the creation of new interpretive and
 20 sensory realities. Artificial intelligence is redefined as a cognitive collaborator
 21 that expands human perception and creativity.

22 Taken together, the three models outline a comparative topology of digital
 23 museology. They demonstrate how artificial intelligence simultaneously
 24 supports material conservation, ethical participation, and aesthetic innovation.
 25 These dimensions can be mapped onto the technical, interpretive, and existential
 26 layers of digital hermeneutics (Capurro, 2010). The technical layer corresponds
 27 to Beijing's focus on infrastructure and automation; the interpretive layer
 28 corresponds to Turin's participatory mediation; the existential layer corresponds
 29 to Cambridge's imaginative exploration of perception and meaning. The
 30 intelligent museum thus emerges as a stratified system in which technology,
 31 ethics, and art are inseparable.

32 Despite their differences, the three contexts reveal several shared challenges.
 33 One is the question of authenticity. In all cases, artificial intelligence destabilizes
 34 traditional notions of originality and authority. Authenticity becomes a relational
 35 quality that arises from interaction rather than a property inherent in the object
 36 (Benjamin, 2019). Another shared issue is transparency. As intelligent systems
 37 mediate interpretation, it becomes necessary to ensure that algorithms remain
 38 accountable and comprehensible. The European model has addressed this
 39 through public documentation of algorithmic design (European Commission,
 40 2021), whereas in the Chinese model, transparency is subordinated to
 41 governance priorities.

42 A third challenge concerns ethics. The use of artificial intelligence to
 43 personalize cultural experience raises questions about surveillance, consent and
 44 data ownership. In both China and Europe, cultural institutions face the dilemma
 45 of balancing innovation with privacy, especially as AI systems collect and
 46 process sensitive behavioural data from visitors (Zhu & Liu, 2025). As Ludovica

1 Russo noted in a recent essay on AI and museums, initiatives such as Historica
 2 imagine AI as an infrastructure for mapping cultural objects, mediating visitor
 3 attention and enabling decentralized participation across borders (Russo, 2025).

4 At the same time, the three models share a common aspiration to
 5 democratize culture through digital means. Whether through Beijing's online
 6 exhibitions, Turin's participatory networks, or Cambridge's immersive
 7 reconstructions, artificial intelligence expands access to knowledge and fosters
 8 new forms of intercultural dialogue. Chinese initiatives such as large-scale
 9 digitization of the Palace Museum's collections and "Digital Dunhuang" have
 10 increased the international visibility of national heritage by making high-
 11 resolution images, 3D models, and immersive experiences accessible to global
 12 audiences. Similarly, European and North American institutions increasingly
 13 collaborate across borders, creating hybrid platforms for shared research, co-
 14 curated exhibitions, and online educational programs (State Council Information
 15 Office of the People's Republic of China, 2021).

16 From a theoretical standpoint, the comparison suggests that artificial
 17 intelligence is producing a new kind of museological consciousness. The
 18 museum is no longer confined to the display of artifacts but extends into a
 19 distributed cognitive environment where knowledge is generated through
 20 interaction among people, technologies, and institutions (Latour, 2005). This
 21 shift requires a rethinking of curatorial authority, which becomes less about
 22 control and more about facilitation. Curators now act as mediators between
 23 algorithmic systems and human audiences, translating data into meaning and
 24 meaning into experience.

25 The global implications of these transformations are profound. The
 26 intelligent museum challenges the traditional separation between technology and
 27 culture, revealing that digital infrastructures are not neutral but carry ideological,
 28 aesthetic, and ethical dimensions. It also calls for new forms of professional
 29 training that combine humanistic sensitivity with computational literacy. As
 30 Ross Parry demonstrated, the success of digital heritage depends not only on
 31 technological sophistication but also on institutions' capacity to cultivate
 32 reflective and inclusive practices (Parry, 2010).

33 In conclusion, the comparative analysis of Beijing, Turin, and Cambridge
 34 illustrates that artificial intelligence is redefining the ontology of the museum. It
 35 transforms conservation into prediction, exhibition into interaction, and curation
 36 into collaboration. Each model contributes a different dimension to the global
 37 evolution of museology: the structural intelligence of Beijing, the ethical
 38 intelligence of Turin, and the imaginative intelligence of Cambridge. Together,
 39 they outline a new paradigm in which the museum becomes an active participant
 40 in the ongoing dialogue between humanity and technology.

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43 **The Future of Museums in the AI Era**

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45 The integration of artificial intelligence into museum practice marks a
 46 decisive shift in the cultural and epistemological functions of these institutions.

1 The museum is evolving from a repository of memory into an intelligent system
2 that generates, organizes, and communicates knowledge through interactive and
3 adaptive technologies. Artificial intelligence introduces a reflexive dimension in
4 which museums become laboratories for exploring cognition, emotion and
5 collective imagination (Cameron & Kenderdine, 2007; Cameron, 2021).

6 The future of museums shaped by artificial intelligence will depend on three
7 interrelated dimensions: cognition, participation, and sustainability. In the
8 cognitive dimension, artificial intelligence will act as a collaborator in research
9 and interpretation. Algorithms capable of analyzing visual, textual, and
10 contextual data will assist curators in identifying patterns and connections that
11 were previously invisible. The process of meaning-making will thus become a
12 hybrid dialogue between human interpretation and machine inference (Capurro,
13 2010). This collaboration will not diminish human creativity; instead, it will
14 expand the interpretive horizon of museums by incorporating computational
15 insights into historical and aesthetic understanding.

16 The participatory dimension concerns the democratization of cultural
17 experience. Intelligent systems can personalize interaction without erasing
18 collective meaning. Emerging AI techniques such as natural language interaction
19 and adaptive, emotionally aware interfaces can be understood as new tools for
20 making museum communication more responsive, furthering the shift from
21 spectatorship to participatory engagement that Giaccardi associates with social
22 media–driven heritage practices (Giaccardi, 2012). However, this transformation
23 also entails ethical responsibility. As Derda and Predescu argue, the use of
24 artificial intelligence in cultural institutions must remain human-centred,
25 ensuring transparency, inclusivity and accessibility. The success of the intelligent
26 museum will depend on how effectively it balances personalization with public
27 accountability (Derda & Predescu, 2025). The success of the intelligent museum
28 will depend on how effectively it balances personalization with public
29 accountability.

30 Sustainability will constitute the third and increasingly urgent dimension of
31 future museology. The environmental impact of digital infrastructures and large-
32 scale computation cannot be ignored. Projects such as I-Muse have demonstrated
33 that artificial intelligence can support sustainable management by optimizing
34 energy use and monitoring environmental conditions (Politecnico di Torino,
35 2023). Future museums will need to adopt “green AI” principles that promote
36 energy efficiency and ethical hardware production while maintaining the
37 creative and educational functions of technology. Sustainability, in this context,
38 refers not only to ecology but also to the long-term cultural and social viability
39 of digital heritage.

40 The ethical dimension will continue to shape the future of the intelligent
41 museum. The questions of privacy, authorship, and authenticity will require
42 continuous reflection and negotiation. Artificial intelligence challenges
43 conventional notions of authority by redistributing agency among curators,
44 visitors, and machines. It also introduces new forms of authorship in which
45 creative responsibility becomes shared. As Benjamin reminds us, technological
46 reproduction alters the “aura” of the artwork, but in the digital age, this alteration

1 can lead to new forms of emotional and cognitive engagement (Benjamin, 2019).
 2 Authenticity in the intelligent museum will not depend on the originality of
 3 objects but on the integrity of the interpretive process that connects them to
 4 audiences.

5 Global collaboration will define the next stage of digital museology. The
 6 networked nature of artificial intelligence facilitates exchanges of data, expertise,
 7 and creativity among institutions across continents. Transnational initiatives
 8 linking China, Europe, and North America are already demonstrating the
 9 potential for shared infrastructures that support cross-cultural understanding.
 10 Chinese digital heritage projects increasingly participate in international
 11 networks, contributing datasets, technical expertise, and immersive content to
 12 global platforms for cultural exchange. At the same time, European and North
 13 American museums experiment with AI-enabled touring exhibitions, joint
 14 digital catalogues, and shared research infrastructures, indicating that the
 15 intelligent museum can become a central platform for global cultural dialogue
 16 (Hawthorne, 2025).

17 From a theoretical perspective, the future of museums will require a
 18 synthesis between technological rationality and humanistic imagination.
 19 Latour's concept of actor networks provides a valuable framework for this
 20 synthesis, showing that meaning arises through the interaction of diverse
 21 agencies rather than through unilateral control (Latour, 2005). As Ross Parry
 22 observes, digital technologies have broadened participation and widened
 23 museums' creative horizons, helping to define a new cultural role for museums
 24 rather than serving efficiency alone (Parry, 2007, p. xii).

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27 **Conclusions**

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29 The findings presented in this article suggest that artificial intelligence is not
 30 simply being “added” to museums as a new layer of technological enhancement.
 31 Rather, it is reconfiguring the museum’s epistemic architecture by shifting how
 32 cultural knowledge is produced, validated, and experienced. Across Beijing,
 33 Turin, and Cambridge, AI appears less as a neutral tool than as an interpretive
 34 infrastructure that redistributes agency across institutions, datasets, interfaces,
 35 and publics. The museum increasingly operates as a dynamic cognitive
 36 ecosystem, where interpretation emerges through relations among human and
 37 nonhuman actors rather than through a one-directional transmission of expert
 38 knowledge (Latour, 2005; Capurro, 2010). This transformation compels a
 39 reframing of digital museology: the decisive issue is not whether AI improves
 40 efficiency, but how it reorganizes authority, authenticity, and cultural meaning
 41 within the museum field.

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The comparative framework developed here clarifies that AI-driven
 museology is evolving along at least three ideal-typical paradigms: infrastructure,
 mediation, and imagination. The Beijing model demonstrates the infrastructural
 paradigm, where AI is integrated into large-scale governance frameworks that
 connect heritage preservation with national digital strategies. In this context,

1 intelligence is primarily institutional and cybernetic, expressed through system-
 2 wide coordination, predictive conservation, and algorithmic management of
 3 access and interpretation. The Turin model illustrates AI as mediation, where
 4 intelligent systems are designed to support participation, transparency, and
 5 human oversight, aligning technological innovation with social responsibility
 6 and sustainability principles (European Commission, 2021; Politecnico di Torino,
 7 2023). The Cambridge case develops AI as imagination, where immersive
 8 reconstruction, simulation, and algorithmic aesthetics reposition the museum as
 9 a laboratory of possible worlds, extending heritage from preservation toward
 10 experiential knowledge and speculative interpretation (Cameron & Kenderdine,
 11 2007; Kenderdine, 2021). Considered together, these paradigms show that the
 12 “intelligent museum” is not a single destination but a plural field of institutional
 13 choices, shaped by political cultures, ethical norms, and epistemological
 14 traditions.

15 These models also highlight a shared tension around authenticity and
 16 interpretive legitimacy. As AI contributes to restoration, narrative adaptation,
 17 and immersive simulation, authenticity becomes less a property of the object and
 18 more a quality of the interpretive process and its transparency. The museum’s
 19 authority is increasingly grounded in the credibility of its workflows: the
 20 visibility of data provenance, the disclosure of algorithmic assumptions, and the
 21 capacity to communicate uncertainty and multiplicity to publics. In this sense,
 22 AI does not merely challenge the museum’s traditional aura, it displaces the
 23 locus of aura from the uniqueness of the artifact to the affective intensity and
 24 relational meaning generated by the encounter, whether physical or digital
 25 (Benjamin, 2019). This shift can enrich cultural experience, but it also increases
 26 institutional responsibility, since algorithmic outputs may harden speculative
 27 reconstructions into seemingly “objective” truths if epistemic humility is not
 28 structurally built into exhibition design.

29 Ethics therefore cannot remain an external constraint applied after
 30 implementation. The analysis suggests that ethical governance must be treated
 31 as part of museological design itself, particularly in relation to personalization,
 32 behavioural data collection, and the risk of turning cultural experience into a
 33 space of implicit surveillance. Visitor-centred systems can deepen engagement
 34 and accessibility, yet they can also introduce asymmetries of power when data
 35 ownership, consent, and algorithmic explainability are insufficiently addressed
 36 (European Commission, 2021; Derda & Predescu, 2025). This is not only a
 37 matter of privacy compliance. It is a question of cultural citizenship: the
 38 museum’s public mission is undermined if visitors cannot understand how their
 39 behaviours are translated into curated pathways, recommendations, or
 40 interpretive hierarchies. For this reason, the intelligent museum should be
 41 conceived as a pedagogical institution not only about heritage, but also about
 42 algorithmic mediation itself, fostering public literacy concerning how AI shapes
 43 perception and meaning.

44 Sustainability further expands the scope of responsibility. The museum
 45 sector increasingly faces a dual obligation: to protect heritage while
 46 acknowledging the environmental footprint of digital infrastructures. AI-enabled

1 optimisation of climate control, energy consumption, and maintenance systems
 2 indicates real potential for reducing ecological costs, but this potential must be
 3 evaluated against the broader material and energetic demands of computation,
 4 storage, and hardware cycles (Politecnico di Torino, 2023). The future of digital
 5 museology will therefore require a more explicit convergence between ethical
 6 AI and green AI principles, in which computational ambition is balanced with
 7 ecological accountability. Sustainability should be understood in a
 8 comprehensive way that includes environmental impact, institutional resilience,
 9 and the long-term integrity of digital heritage preservation.

10 Several limitations of this study should also be acknowledged. The
 11 comparative design, while conceptually productive, is based on a limited number
 12 of emblematic cases rather than on a broad quantitative mapping of global
 13 practices. In addition, the rapid evolution of generative AI, multimodal interfaces,
 14 and immersive infrastructures means that institutional models are likely to
 15 change faster than scholarly frameworks. Future research should therefore
 16 extend this comparative topology by including additional regions and
 17 governance regimes, with particular attention to the Global South, where
 18 infrastructural constraints and cultural priorities may generate alternative
 19 paradigms of intelligent museology. Further work is also needed on evaluation
 20 methods capable of measuring not only usability or satisfaction, but interpretive
 21 pluralism, perceived legitimacy, and ethical trust. Longitudinal studies would be
 22 especially valuable for understanding how repeated exposure to AI-mediated
 23 heritage reshapes learning, memory, and cultural identity over time.

24 Ultimately, the central implication of this article is that the success of AI in
 25 museums will not be determined by the sophistication of algorithms alone. It will
 26 depend on the institutional capacity to integrate technological innovation with
 27 epistemic transparency, ethical governance, participatory legitimacy, and
 28 ecological responsibility. The museum of the AI era can preserve the past while
 29 generating new interpretive futures, but only if it treats intelligence as a cultural
 30 relation rather than as an automated substitute for human meaning-making. The
 31 intelligent museum, in this sense, is best understood as a continuously negotiated
 32 interface between memory and prediction, where the renewal of cultural
 33 meaning remains the primary criterion of progress.

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