

Developing a Digital Illustration Curriculum Based on A/R/Tography: Integrating AI-Generated Art and Chinese-Spanish Heritage Patterns for Cross-Cultural Education

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Abstract—A/R/Tography, as a research framework that integrates artistic creation (A), research (R), and teaching (T), provides new possibilities for digital art education. This study explores the application of Generative AI (AIGC) in heritage pattern design by integrating AI-generated art with A/R/Tography theory to develop a structured digital illustration curriculum. The course framework comprises symbolic deconstruction, AI-assisted pattern generation, digital reconfiguration, and cross-cultural adaptation, enabling students to engage critically with Chinese and Spanish heritage motifs, such as Sanxingdui totems, Forbidden City cloud and dragon patterns, Alhambra geometric structures, and Arabesque floral ornaments. By closely working with AI Agents, students will analyze and generate heritage symbols and optimize AI-created patterns through manual double-checking and cultural adaptation, ensuring aesthetic authenticity and global applicability. Findings indicate that AI enhances efficiency, accessibility, and innovation in heritage design education, whereas human intervention remains essential for cultural accuracy and creative extension. This research demonstrates that AI+A/R/Tography not only enriches artistic pedagogy, but also fosters cross-cultural dialogue, interdisciplinary learning, and the global dissemination of heritage art in the era of digitalization.

Keywords—A/R/Tography, Generative AI, digital illustration, heritage patterns, cross-cultural adaptation, visual arts education

I. INTRODUCTION

In the era of globalization and rapid technological advancement, art education encounters unprecedented opportunities and complex challenges. A/R/Tography—an innovative methodology that integrates artistic creation, research, and teaching—offers a multidimensional framework that redefines contemporary art education. By emphasizing the dynamic interplay among the roles of artist, researcher, and educator, this approach fosters a learning experience that extends beyond traditional classroom settings. Engaging in artistic practice enables learners to gain a deeper understanding of the cultural and social significance of the arts [1]. Within this theoretical context, China's Intangible Cultural Heritage (ICH)—including Sanxingdui patterns, traditional motifs from the Forbidden City, and woodblock New Year paintings, has become a valuable resource for creative exploration in modern design education. Likewise, Spanish Moorish art, particularly the geometric patterns and arabesque motifs of the Alhambra Palace in Granada, is

recognized globally for its perfect symmetry and aesthetic richness, making it a vital component of art education worldwide [2].

With the rapid advancement of Artificial Intelligence (AI), generative AI technologies, such as stable diffusion, MidJourney, DALL-E, and Deepseek, are fundamentally transforming artistic creation and education. These tools not only produce highly complex and diverse patterns but also enable the digital reinterpretation of traditional art forms through algorithmic processes. This technological innovation broadens students' creative possibilities and opens new avenues for modernizing and globally disseminating conventional ICH motifs [3]. Despite these advancements, current research on ICH pattern education rarely examines how generative AI can be effectively integrated with traditional artistic heritage, nor does it adequately address how cross-cultural design perspectives can contribute to the international promotion of ICH in art education.

Against this backdrop, this study addresses three key research questions: (1) How can digital illustration courses on ICH patterns be developed using an A/R/Tography-based framework? (2) How can the integration of generative AI technology with A/R/Tography-informed pedagogical approaches enhance the modernization of ICH pattern education? (3) How can curriculum design improve the adaptability and global impact of ICH patterns in cross-cultural contexts? These inquiries aim to broaden the scope of art education, promote cultural exchange and resonance in a globalized world, and offer interdisciplinary solutions to the challenges facing the preservation and evolution of traditional artistic heritage.

This study aims to construct a multidimensional model tailored to contemporary art education by integrating A/R/Tography with generative AI technologies to enhance ICH's digital education and global communication. It conceptualizes AI agents as "Mediators" of cultural motifs and categorizes pedagogical approaches into learner-centered, teacher-centered, and AI-agent-centered models. This research further proposes an AI+ART application framework to support the transformation of traditional motifs into visual languages and expressions that align with global aesthetic sensibilities. Within a digital illustration curriculum focused on ICH motifs, AI-generated patterns not only deepen students' understanding of cultural heritage but also cultivate

their international creative competencies through cross-cultural design practices. This study explores the potential of generative AI in cultural heritage preservation and artistic innovation, presenting new avenues for interdisciplinary exchange between Eastern and Western traditional arts in the digital age.

II. LITERATURE REVIEW

A. A/R/Tography: Theory and Development

A/R/Tography is an interdisciplinary methodology that examines the dynamic interplay among the roles of artist, researcher, and teacher, offering a multidimensional framework for art education [2]. Its core philosophy focuses on fostering students' cultural awareness and interdisciplinary thinking through artistic creation and critical reflection, while also uncovering the historical and contemporary significance of visual arts practices [4].

In recent years, A/R/Tography has been increasingly validated across cultural, historical, artistic, and educational research, demonstrating its effectiveness in helping students gain a deeper understanding of the historical context and contemporary relevance of cultural motifs through artistic creation and self-reflection. However, its successful application in technology-assisted teaching remains limited, particularly in the integration of generative AI [5]. A significant challenge in digital art education lies in effectively combining AI technologies with A/R/Tography to fully harness its potential for artistic creation and cultural critique [6].

Emerging studies have begun to explore the intersection of materiality and digital practices, with research on "Material and Digital A/R/Tography Exploration" showing that digital tools offer students greater flexibility in transforming traditional cultural symbols into contemporary artistic expressions [7]. In China, A/R/Tography has been increasingly integrated with local cultural traditions, particularly within the context of ICH education, highlighting its potential in this field. The continued development of digital A/R/Tography underscores its growing relevance in modern technological settings [6]. However, while these studies contribute to the theoretical expansion of the framework, systematic support for its practical application remains limited. Future research should focus on integrating generative AI technologies into A/R/Tography-based education, creating a model that merges cultural heritage preservation with technological innovation—ultimately offering promising pathways for safeguarding and globally disseminating ICH [8].

B. Education and Digitalization of Intangible Cultural Heritage Patterns

ICH patterns carry deep cultural significance, with examples such as the Sanxingdui motifs reflecting the religious beliefs of ancient Shu, the traditional designs of the Forbidden City embodying the Chinese spirit, and woodblock New Year pictures conveying folk customs through educational and narrative elements. These patterns offer valuable opportunities for enhanced cultural transmission through digital technologies [9, 10]. However, current research often focuses on isolated symbols and lacks comprehensive historical analysis and cross-cultural

comparison, which limits their integration into art education and international dialogue [11]. Although advancements in digital museums and augmented reality technologies have emerged, the documentation and presentation of ICH patterns still lack standardized protocols, hindering public understanding of their cultural importance [12]. For example, although Chinese New Year pictures and the geometric patterns of the Alhambra in Spain each exhibit unique cultural characteristics, the lack of comparative studies and robust international communication frameworks hampers educational development and the global recognition of ICHs [13].

C. Educational Potential of AI-Generated Art

Generative AI has emerged as a transformative force in visual creation, fundamentally reshaping the paradigms of art education [8]. As a creative tool, it not only expands artistic boundaries but also promotes critical reflection. Visual A/R/Tography, which explores cultural narratives and identities through artistic practice, aligns naturally with generative AI, making their convergence increasingly relevant [14]. Currently, AI functions as an educational assistant in the student-teacher-agent-centered instructional model. However, scholars caution against viewing AI merely as a creative aid, arguing that this limited perspective risks neglecting the post-human, symbiotic relationship between humans and machines that is increasingly shaping educational environments [15]. This perspective indicates that AI has reshaped not only the ethics of learning but also creative identity and autonomy [16], as seen in the AI-CTSAM model, which integrates social learning theories to foster student-centered innovation, yet still prioritizes cognitive outcomes over cultural considerations [17].

These analyses underscore the need to critically redefine the role of AI in art education—not merely as a tool for efficiency or innovation, but as a medium for cross-cultural translation and symbolic interpretation. In response, this study positions AI as a cultural translation agent within the framework of A/R/Tography, proposing a model in which AI contributes to both the formation and transnational dissemination of ICH. This perspective aligns with the view of Holmes *et al.* (2019), who advocate for the reformulation of art education curricula through AI integration—not only to enhance teaching effectiveness but also to address the social dimensions of AI, emphasizing the importance of balancing technological advancement with critical cultural awareness [18].

Within the framework of this study, AI enhances the interactivity and creativity of art courses by providing real-time feedback and creative suggestions. This approach establishes a foundation for integrating generative AI into art education and advancing the digital transformation of A/R/Tography. For example, AI-generated variations of traditional woodblock New Year pictures have been shown to boost student creativity and make cultural heritage more engaging and accessible to younger audiences [19]. Despite these benefits, significant challenges remain, as current research lacks effective strategies for integrating generative AI with A/R/Tographic practices [20]. Future studies should investigate how AI can support cultural interpretation and cross-cultural expression, offering both theoretical insights

and practical tools to strengthen global art education.

III. METHODOLOGY

This study is grounded in A/R/Tography theory and integrates generative AI technology with the cross-cultural dissemination of ICH motifs. A comprehensive framework was developed by combining the concepts of “AI Translation Agent” and “Cultural Translation Code” to create a digital learning system that aligns with contemporary art education and cultural transmission. The following sections provide an in-depth exploration of the theoretical framework, technical model design, and course development.

A. Theoretical Framework

The theoretical model constructed in this study (see Fig. 1) integrates three teaching dimensions: The bottom layer involves the generation and adaptation of ICH patterns using AI tools. The middle layer represents the collaborative mechanism of the three roles defined in the A/R/Tography framework—Artist, Researcher, and Teacher. The top layer emphasizes illustrative design applications, such as branding, UI design, digital multimedia, and other cross-cultural implementations. This three-tiered architecture not only illustrates the transformation of ICH patterns into digital assets but also highlights AI’s dual role as a technical tool and cultural intermediary within educational settings, promoting the deep integration of generative AI into art education.

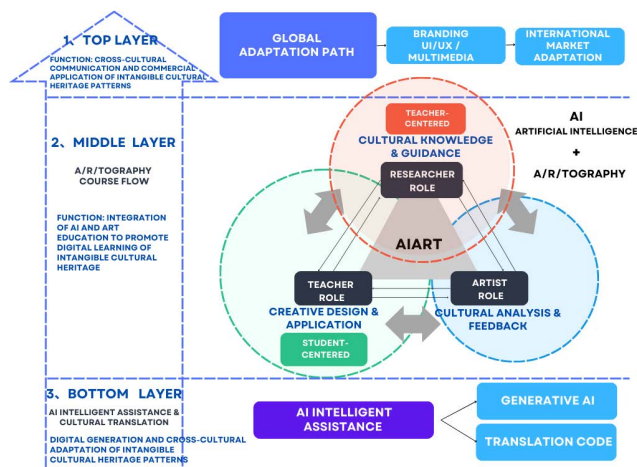


Fig. 1. A/R/Tography-AI integrated theoretical model for cross-cultural illustration education.

Bottom Layer: This layer focuses on AI-assisted generation and cultural translation codes. Generative AI tools, such as MidJourney and DALL-E, are employed to create ICH patterns, which are then refined through symbolic decoding, visual recomposition, and contextual adaptation to improve their relevance and applicability across diverse cultures [21]. The primary goal of this layer is to optimize AI-generated ICH motifs to ensure that they align with the aesthetic preferences and market demands of diverse cultural contexts.

Middle Layer: Grounded in A/R/Tography theory, this level establishes a tripartite collaborative model involving educators, researchers, and artists [22]. Educators offer critical cultural contexts and pedagogical guidance, researchers refine the logic and processes behind AI generation, and artists reinterpret AI-generated materials into

original creative works. This layer underscores the methodological importance of integrating AI into art education, enabling students to gain a deeper and more holistic understanding of ICH through AI-enhanced creative processes.

Top layer: This layer focuses on three main applications—branding, UI/UX design, and digital multimedia—aimed at the international dissemination of ICH motifs. AI-generated ICH patterns are applied in various contexts, including packaging design, interactive user interfaces (UI), non-fungible tokens (NFTs), and augmented and virtual reality (AR/VR) experiences, all of which support global cultural communication. The goal of this layer is not only to broaden educational horizons through the creative use of AI-generated ICH patterns but also to establish them as a new medium for cultural transmission in the global market.

This model overcomes the limitations of traditional ICH art education by introducing AI as a medium for cultural translation. Integrating AI with A/R/Tography creates a dynamic cycle of generation, application, and dissemination, providing a digital learning pathway that enhances the global visibility, relevance, and adaptability of ICH within contemporary design contexts.

B. Technical Model

To facilitate the deep integration of generative AI and A/R/Tography theory in art education, this study introduces a systematic, multi-layered interactive model comprising four key layers: Input, AI-assisted translation, Teaching Flow, and Output. Designed around the cultural characteristics and educational significance of ICH motifs, the model aims to cultivate students’ cultural identity and creative abilities through the synergy of AI technology and art education. AI in education (AIED) enhances this process by enabling personalized learning, supporting educators, and optimizing administrative tasks, thereby contributing to the overall advancement of educational practices [3].

As shown in Fig. 2, the technical model developed in this study comprises four interrelated levels representing the complete teaching and production processes. The model begins with the input of ICH theme data and incorporates AI-assisted generation, collaborative teaching activities, and practical application outputs. Each level aligns with a specific stage of the digital illustration course, facilitating the transformation of traditional cultural and artistic patterns into visual assets adaptable to global contexts. The following sections provide a detailed explanation of each layer within the model:

1) Input layer

This layer builds a comprehensive database of ICH motifs, featuring several cultural symbols, such as Sanxingdui ancient Shu patterns, Mianzhu woodblock New Year paintings, and geometric motifs from the Alhambra Palace. Using deep learning algorithms, this method analyzes the historical context, visual structures, and cultural meanings embedded in these patterns to ensure systematic organization while preserving data diversity. This layer serves as a precise cultural input foundation for generating contextually rich and accurate AI-generated outputs.

2) AI-assisted translation layer

This layer employs AI tools such as MidJourney, DALL-

E, and Stable Diffusion to generate ICH patterns that align with both market trends and cultural expectations. The AI Translation Agent refines these AI-generated motifs through symbolic decoding, style transfer, and contextual adaptation to ensure visual coherence with the cultural contexts of target markets. Simultaneously, AI engages in cross-cultural style learning by extracting design logic, color preferences, and symbolic systems from various cultures to preserve the authenticity of ICH motifs while enhancing their adaptability to global audiences [1].

3) Teaching flow layer

This level develops courses grounded in A/R/Tography theory, emphasizing collaborative engagement among educators, researchers, students, and the AI Translation Agent. Educators offer cultural guidance [3], researchers enhance the logic behind AI generation, and students reinterpret AI-generated motifs using creative tools such as Procreate and Photoshop to produce original artwork. This collaborative process fosters students' cross-cultural artistic expression and deepens their understanding of cultural heritage through both technological and artistic lenses [19].

4) Output layer

This level centers on the application of AI-generated ICH motifs in branding, UIs, and digital media. These creative outputs are distributed globally through platforms such as NFTs, AR/VR immersive experiences, and digital exhibitions, and are evaluated based on criteria including creativity, technical execution, and cultural adaptability to ensure their relevance and marketability in diverse cultural contexts [23].

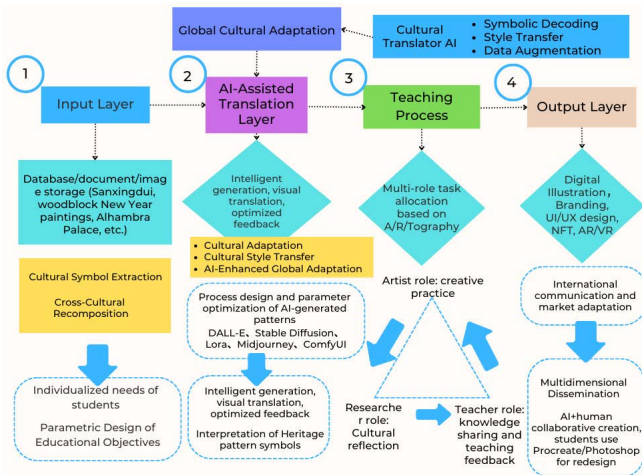


Fig. 2. Technical model for AI-assisted cross-cultural pattern generation.

The four-layer technical model converts the theoretical framework into practical application. The input layer establishes a comprehensive database of pattern symbols from the Chinese and Spanish ICH. The AI-assisted translation layer uses tools such as Stable Diffusion and LoRA for stylized AI agent training, enabling style transfer and cultural adaptation of patterns and designs from both countries. The teaching process layer emphasizes human-machine collaboration, where students use design software to refine and enhance AI-generated outputs. Finally, the output layer applies the completed illustrations to real-world contexts such as brand promotion, UI/UX design, and AR/VR experiences, thus supporting the global dissemination of cultural heritage.

C. Course Design Process

Grounded in A/R/Tography theory and generative AI, this study develops a cross-cultural digital illustration curriculum consisting of four core modules: Traditional Learning, AI-Generated Pattern Creation, Digital Illustration, and Cross-Cultural Design Adaptation. By integrating the AI Translation Agent, the course establishes a comprehensive learning pathway from cultural understanding to international dissemination. This curriculum empowers students to develop a visual language for ICH motifs while leveraging AI to enhance their global appeal, fostering innovative reinterpretations of traditional art in the digital age.

Students begin their exploration by analyzing cultural symbols such as Sanxingdui motifs, Mianzhu woodblock prints, clouds and phoenix patterns from Beijing's Forbidden City, and Moorish designs from the Alhambra Palace [22]. Through semiotic analysis and cross-cultural comparison, these scholars investigated the symbolic meanings and visual languages embedded in these traditions [18]. Using AI tools like MidJourney, DALL-E, and Stable Diffusion, students generate diverse variations of ICH motifs, which are then refined through symbolic decoding, style transfer, and contextual adaptation to ensure market alignment with contemporary market and cultural expectations.

Subsequently, students enhance AI-generated patterns by adjusting colors, refining compositions, and arranging motifs using tools like Procreate and Photoshop to enhance cultural storytelling and artistic expression [23]. In the final phase, selected works are adapted for market applications, including branding, UI interfaces, NFTs, and AR/VR experiences. The evaluation criteria emphasize creativity, technical execution, and cultural adaptability to assess their potential for global dissemination [24]. This curriculum not only cultivates digital artistic skills but also provides a technological framework for the global adaptation and promotion of ICH Patterns.

IV. COURSE CASE STUDIES AND APPLICATIONS

The evolution of AI-generated content (AIGC) has created a new pathway for the digital translation and cross-cultural dissemination of ICH [25]. By applying A/R/Tography theory and incorporating AIGC technology, this study explores AI's role in generating traditional motifs, facilitating cross-cultural adaptation, and enabling creative reinterpretation. It develops a pedagogical model that integrates cross-cultural comparison, AI-driven innovation, and adaptation to international markets, while also introducing the AI Translation Agent to enhance the global adaptability of ICH motifs within the design marketplace.

A. AIGC Principles: Theoretical Foundations and Methods of AI-Generated Art

The significant advancement of AIGC technology has facilitated the intelligent recreation of ICH motifs, thus supporting their adaptation to contemporary visual design requirements. Globally, technology companies such as OpenAI and Google have extensively implemented AIGC for image generation, while Chinese enterprises such as Baidu, Alibaba, and Tencent continue to advance AI applications in artistic design. AIGC incorporates deep learning algorithms, style transfer, and symbolic recognition technologies to

achieve the digital analysis and recreation of traditional motifs [26]. The AIGC-based ICH motif generation process in this course comprises the following four stages:

Data Input Layer—A multicultural ICH symbol database is constructed, incorporating Sanxingdui totems, Forbidden City decorative motifs, Mianzhu woodblock New Year paintings, and Alhambra Moorish geometric patterns from Spain, to ensure both data diversity and cultural authenticity.

AI Training Layer—This layer uses computer vision and deep learning technologies to analyze the composition logic, color schemes, and cultural symbolism of ICH motifs, thereby establishing a style learning model.

AI Generation Layer—Deploys MidJourney, DALL-E, and Stable Diffusion to generate variations of the ICH motif that align with modern visual aesthetics while refining their visual styles.

Optimization and Feedback Layer—This layer combines human intervention with the AI Translation Agent to fine-tune color schemes, compositions, and cultural adaptability, ensuring that AI-generated motifs align with the dissemination needs of target markets (Fig. 3).

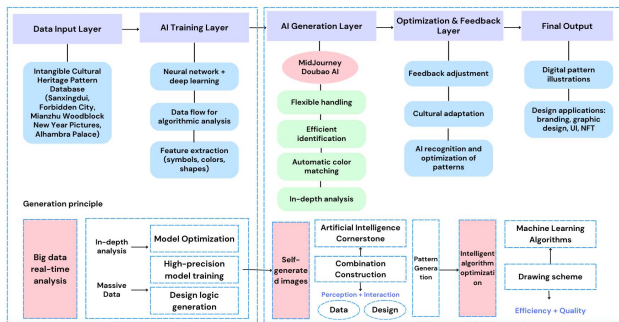


Fig. 3. AIGC-based four-stage process for generating ICH motifs.

MidJourney, Stable Diffusion, Doubao AI, and DALL-E are prominent AIGC tools, each offering distinct capabilities. In particular, stable diffusion demonstrates strong effectiveness in style adaptation and color optimization, enabling broad applications across branding, UI design, and digital media. These capabilities facilitate the digitization of ICH motifs and help bridge traditional art with contemporary design [10]. However, AIGC faces significant challenges related to creativity, cultural authenticity, and ethical considerations [27]. To ensure the cultural appropriateness of AI-generated patterns, it is essential to balance automated processes with artistic innovation. Furthermore, optimizing symbolic interpretation, stylistic adaptation, and marketing communication is crucial to supporting the sustainable cross-cultural application of these motifs.

Fig. 3 summarizes the generation process and outlines four consecutive stages: beginning with ICH data input and style learning, progressing to AI-assisted generation, and then advancing to manual refinement and feedback. This process ensures not only the technical feasibility of visual output but also cultural sensitivity.

This diagram illustrates the four essential layers of the AIGC-driven generation process implemented in the course. The data-input layer establishes a comprehensive database of cultural themes, which forms the foundation for the generation process. The AI training layer uses deep learning and computer vision to analyze composition, color schemes,

and symbolism. The AI-generation layer then employs tools such as MidJourney and Stable Diffusion to produce variations in the themes, aligning them with contemporary visual aesthetics. Lastly, the optimization and feedback layer incorporates human intervention to refine color, form, and cultural consistency, ensuring that the final output meets both design standards and symbolic authenticity.

B. Pattern Symbol Extraction Method: Digital Analysis of ICH Motifs

The patterns of ICH serve as both visual heritage and carriers of cultural significance. Within the AIGC framework, a key challenge lies in preserving authenticity while ensuring cross-cultural adaptability. This study tackles this issue through a digital semiotic analysis framework, which is structured into three stages: symbol decomposition, style classification, and cross-cultural adaptation, all aimed at meeting both cultural and visual requirements.

1) Symbolic decomposition

This research focused on deconstructing ICH motifs to identify core visual elements and develop an AI recognition system using computer technology. It converts key motifs into structured shapes and color combinations, such as Sanxingdui's Eye Totem, the Forbidden City's Cloud-Dragon-Phoenix, Woodblock New Year Deities, and Alhambra's geometric patterns [21]. By using Computer Vision and Deep Learning techniques, the researchers analyzed the deeper meanings of these symbols and created an AI training database aimed at enhancing cultural accuracy.

2) Style classification

The style classification system ensures that AI-generated motifs adhere to cultural logic by organizing them into morphological categories, such as geometric abstraction (e.g., Alhambra and Arabic motifs), symbolic totems (e.g., Sanxingdui and Woodblock prints), and decorative patterns (e.g., Forbidden City dragon-phoenix motifs). Additionally, the color classification system examines the contrast between the high-saturation tones in Chinese woodblock prints and the low-saturation blue-green hues characteristics of Moorish designs [28].

3) Cross-cultural adaptation

To enhance the cross-cultural adaptability of ICH patterns, this study refines the three core functions of the "AI translation agent." Visual translation directs AI to adjust colors and compositions based on an analysis of the aesthetic preferences of target markets. Symbol recombination aligns traditional motifs with brand requirements, transforming original compositions into arrangements that better reflect Western visual rhythms. Context optimization through user feedback and machine learning enables AI to gradually adapt to the expression habits of different cultural contexts, thereby improving the accuracy and cultural resonance of cross-cultural outputs.

C. Course Case Study: AI Generation + Student Creation + Cross-Cultural Adaptation

1) AI-generated cross-cultural patterns based on ICH motifs

In the initial stage of the course, AI tools such as

MidJourney and Doubao AI will be used to generate a foundational dataset of ICH patterns while exploring cross-cultural possibilities. This process follows a structured approach—symbol decomposition, style classification, and cultural adaptation [17]—to ensure that the patterns maintain cultural authenticity and align with global design aesthetics. AI-generated patterns also allow students to observe style differences across cultures, enhancing their understanding and creative application of ICH motifs.

In the practical phase, students collaboratively designed AI prompts under tutor guidance, using tools like Stable Diffusion and MidJourney to generate culturally relevant visuals based on inputs such as “Sanxingdui bronze totem + gold texture + flat vector style”. These prompts were structured across three layers: (1) cultural keywords (e.g., Sanxingdui, Alhambra); (2) stylistic direction, reflecting student preferences (e.g., bright colors, brush texture); and (3) design purpose (e.g., book cover, packaging). Additionally, students may train models, such as LoRA in Stable Diffusion, to develop cross-cultural visual illustration styles.

The output images were evaluated using a scoring scale that considered cultural accuracy, compositional balance, and visual originality. There are two stages of human supervision: First, the tutor assesses the AI-generated images for cultural appropriateness and visual fidelity; Second, students use digital tools (e.g., Procreate, Photoshop) to optimize structures, adjust colors, and recombine elements. This process ensures that AI-generated visuals are both aesthetically pleasing and aligned with cultural narratives and educational goals.

a) First generation: Single-cultural ICH motifs

The course practice follows a three-tier structural approach: (1) cultural keywords, such as “Sanxingdui pattern,” “Alhambra geometric pattern,” or “Palace Museum pattern”; (2) Style references, like “vector style,” “bright colors,” or “flat illustration”; and (3) application scenarios, including “brand,” “poster,” or “bookbinding design.” For example, a student might create a prompt such as “Generate Sanxingdui Divine Bird + vector flat style + red-gold tone for silk packaging.” Through classroom discussions and a comparative analysis with traditional patterns, instructors help students refine these prompts. The following example demonstrates a teaching case presented during class.

In this step, students guided by instructors input AI prompts emphasizing the Chinese ICH elements. Example prompt:

“Generate an illustration in the style of Chinese ICH, incorporating Sanxingdui totems (eyes, sun, divine bird), the Forbidden City’s cloud and dragon-phoenix motifs, and the storytelling and folk traditions of woodblock New Year paintings”.

The AI-generated patterns serve as visual references for students’ semiotic analysis, emphasizing the distinctive artistic language of Chinese traditional art. Instructors facilitate discussions on cultural contexts, encouraging students to explore how symbols communicate social and historical narratives while fostering critical and reflective reinterpretation of these visual elements. Fig. 4 showcases examples of AI-generated motifs.

This image illustrates the initial phase of AI-assisted pattern generation, focusing on ICH patterns from a single

cultural source and providing a visual foundation for symbolic analysis and classroom discussion.



Fig. 4. AI-generated Chinese patterns based on Sanxingdui and forbidden city motifs.

b) Second generation: Integrating cross-cultural elements

Within the A/R/Tography framework, students explore cultural interactions by modifying AI prompts under the guidance of instructors. The instructors help students incorporate Spanish cultural elements into their designs by encouraging them to add features such as *Alhambra’s geometric patterns (symmetry, golden ratio)* and *Arabic motifs (vine, floral ornamentation)*. This process enables students to achieve cross-cultural fusion by building on previously generated ICH motifs to create visually rich and culturally integrated designs.

This phase aims to deepen students’ understanding of the similarities and differences between design logic and symbolic meanings across cultures. Using style transfer techniques, AI blends the fluidity of traditional Chinese cloud patterns with the geometric precision of Moorish designs, highlighting cultural elements’ adaptability across diverse visual styles. The AI-generated outcomes highlight the potential to integrate patterns from different cultures, offering a tangible example of cross-cultural fusion (see Fig. 5).



Fig. 5. Cross-cultural patterns generated by AI incorporating Chinese and Spanish motifs.

This example demonstrates how AI can be used to integrate design elements from China and Spain, helping students explore cross-cultural symbolic meanings through guided prompt engineering.

c) Third generation: Cultural comparison and reinterpretation

In this phase, instructors guide students through a comprehensive analysis of AI-generated Sino-Spanish fusion

patterns, followed by creative reinterpretation using Photoshop and Procreate. Students begin by examining the compositional principles, symbolic representations, and color schemes in the AI-generated motifs, identifying both the strengths and limitations of AI in cross-cultural adaptation. They then incorporate their artistic perspectives by refining AI-generated designs to better align with contemporary branding aesthetics. Within the A/R/Tography framework, students critically reflect on AI's role in artistic creation and evaluate its effectiveness in cultural dissemination, assessing its adaptability across diverse visual markets (see Fig. 6).

Main Prompt: “Create an intricate cultural illustration blending Chinese and Spanish heritage patterns, showcasing their design logic, symbolism, and artistic styles.”

Chinese Elements: Sanxingdui Totems (Eye, Sun, Divine Bird) representing religious and spiritual significance; **Forbidden City Motifs** (Cloud, Dragon, Phoenix) symbolizing imperial grandeur and auspicious meanings; **Woodblock New Year Prints** reflecting narrative-driven folk art.

Spanish Elements: Alhambra Geometric Patterns—Symmetrical, mathematically precise designs; **Arabesque Floral Ornaments:** Nature-inspired abstract patterns.

Style: Flat vector with detailed ornamentation and vibrant yet balanced colors. **Focus:** Chinese storytelling motifs with **Spanish geometric precision.** **Output:** Integrated pattern illustration, motif reinterpretation, and cultural annotation.



Fig. 6. Student-reinterpreted pattern derived from AI-generated output.

In the final stage, students use design software to refine the visual elements generated by AI, enhancing symbolic expression, cultural relevance, and market viability.

2) Student creations: AI+ART integrated teaching process

Students digitally enhance AI-generated patterns to enrich their cultural meaning and market relevance. Although AI outputs provide foundational symbolic structures, they require manual adjustments to colors, layouts, and symbolic elements. Enhancement strategies include refining color schemes to match target market preferences, reconfiguring compositions to align with contemporary design trends, and emphasizing underrepresented cultural symbols to ensure authenticity and visual appeal.

The refined works were applied to the branding, UI design,

and market adaptation tests [27]. For cultural product packaging, students combine Sanxingdui totems, Forbidden City cloud motifs, and woodblock New Year paintings with Alhambra geometric patterns and Arabic floral ornaments to design tea boxes and silk scarf packaging that appeal to international markets (see Fig. 7). In UI interface design, AI-generated fusion patterns are integrated into travel app backgrounds and icon decorations, optimized using tools like Figma or Adobe XD to ensure a balance between aesthetics and user experience (see Fig. 8).

This methodology improves the efficiency of ICH pattern reinterpretation while broadening AI's role within A/R/Tography theory, positioning AI as more than just a generative tool—it becomes an integrated catalyst for creation, research, and teaching. The interactive dynamics between instructors, students, and AI establish AI as an intelligent cultural translator, enabling the cross-cultural reinvention of traditional motifs and opening up new possibilities for art education and global dissemination.



Fig. 7. Cross-cultural packaging design using AI-enhanced traditional patterns: Student applications in tea box and scarf branding.



Fig. 8. Integrating AI-generated cultural patterns into travel App UI design: Enhancing aesthetics and user experience.

3) AI output evaluation, culture-fit marketing applications, and human oversight and reproduction processes

The evaluation of AI-generated outcomes follows a three-dimensional assessment framework: (1) cultural accuracy, which assesses how well the patterns adhere to historical precedents and symbolism; (2) Design quality, focusing on visual composition, balance, and color harmony; and (3) Marketability, which evaluates the suitability of the designs for their intended applications. Students and faculty conduct these evaluations through peer reviews, visual analysis, and reflective commentary, often referencing original ICH motif archives to ensure authenticity and relevance.

This study highlights AI's integral role within the A/R/Tography framework through an experimental teaching model that combines “AI generation, student creation, and market adaptation.” AI acts as an intelligent cultural interpreter, deepening students' understanding of intangible cultural patterns while bridging traditional and contemporary design elements [12]. In a collaborative framework that involves instructor guidance, researcher optimization of

generative algorithms, and student reinterpretation, AI transcends its role as a supplementary tool to become a crucial medium for cross-cultural dialogue. The market analysis shows that AI-enhanced ICH patterns achieve greater cross-cultural acceptance, which aligns with modern design's pursuit of innovative traditional art expressions.

Human oversight is a consistent aspect of the AI-assisted design process. Instructors offer cultural contexts to preserve the symbolic integrity of patterns, while students use digital platforms such as Photoshop and Procreate to critically evaluate and refine AI-generated outputs. This collaborative approach transforms AI from a mere generative tool to an instructional co-creator, enhancing cultural awareness and design expression. Review discussions are integral to the process, ensuring the correction or rejection of inaccurate or misleading outputs.

V. SURVEY RESULTS AND ANALYSIS

This pilot study employed a quasi-experimental design to evaluate the pedagogical effectiveness of integrating A/R/Tography with Generative AI (AIGC) in a digital illustration curriculum. Two elective classes ($n = 40$ each) at a university were randomly assigned to either (a) a **traditional instruction** arm—leveraging conventional art-history lectures, manual critique, and standard design software—or (b) an **AIGC-based instruction** arm that incorporated AI translation agents and prompt engineering within an A/R/Tographic framework. Approximately 70 % of the participants in both cohorts were art design majors, ensuring the relevance of the subject background. All 80 questionnaires were completed, de-duplicated, and cleaned for analysis.

A. Descriptive Statistics

Table 1. Mean / SD for key outcomes

Variable	Group	Mean	SD
Cross-Cultural Understanding	Traditional	4.48	1.34
Cross-Cultural Understanding	AIGC + A/R/Tography	5.00	1.54
Creative Design Ability	Traditional	4.93	1.19
Creative Design Ability	AIGC + A/R/Tography	5.00	1.52

As shown in Table 1, the AIGC-integrated cohort gains higher mean ratings in cross-cultural understanding ($M = 5.00$ vs. 4.48) and creative design ability ($M = 5.00$ vs. 4.93) compared to the traditional group. Although our sample size was small and only a small-scale experimental test, the comparative trends of the data curves and qualitative classroom insights provide some quantitative research support for the teaching advantages of AIGC embedded in the A/R/Tography framework.

B. Distribution Comparisons

Figs. 9–11 display the full response distributions (percentage of respondents) for Q6 (Cross-Cultural Understanding), Q7 (Creative Design Ability), Q8 (“Preferred Design Method”), Q9 (“Future Practice Helpfulness”) and Q10 (“Future Impact Perception (A/R/Tography-AIGC integration vs Traditional Pedagogy)”).

1) Q6. Cross-Cultural Understanding (Fig. 9).

The traditional cohort peaked at “4” (Neutral, 45 %), whereas the AIGC cohort’s modal response shifted rightward to “6” (Agree, 30 %), indicating a 20 % increase at ≥ 5 .

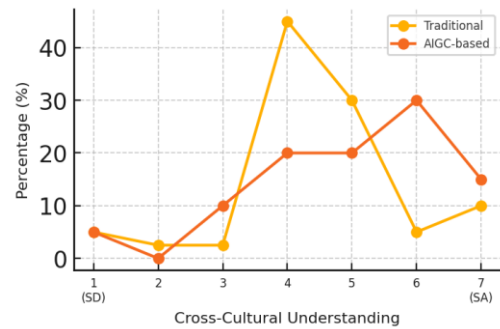


Fig. 9. Cross-cultural understanding distribution.

2) Q7. Creative Design Ability (Fig. 10).

It reveals that while 60 % of the control-rated creative-ability gains ≥ 5 , 62.5 % of the AIGC group did so, with a notable peak at “6” (30 % vs. 17.5 %).

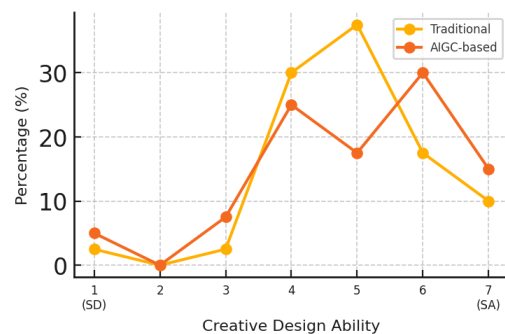


Fig. 10. Creative design ability distribution.

3) Q8. Preferred Design Method.

This result demonstrates that 20 % of AIGC-trained students judged AI-generated methods “most suitable” for cross-cultural motif design, compared to only 7.5 % under traditional instruction—another right shift in preference.

4) Q9. Future Practice Helpfulness (Fig. 11)

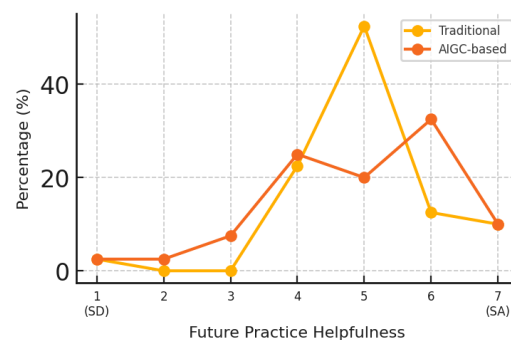


Fig. 11. Future practice helpfulness distribution.

This illustrates that only 12.5 % of the traditional cohort rated the course as “6” (Substantially Helpful), whereas 32.5 % of the AIGC-integrated cohort did so. Moreover, 82.5 % of AIGC-based students rated ≥ 5 compared to 75 % in the control group. This pronounced rightward distribution shift indicates that students exposed to AIGC + A/R/Tography pedagogies perceive greater applicability of course concepts to future design and learning practice.

5) **Q10. Future Impact Perception** (A/R/Tography-AIGC integration vs Traditional Pedagogy)

It illustrates that 32.5 % of the AIGC cohort predicted the “transformative” impact of AIGC+A/R/Tography on their professional practice (vs. 30 %), with fewer “uncertain” responses (7.5 % vs. 10 %).

C. Limitations

Although this pilot ($n = 40$ per arm) exhibits consistent rightward shifts favoring A/R/Tography–AIGC integration across all five measures—suggesting enhanced cross-cultural understanding, creative design ability, methodological preference, future practice helpfulness, and perceived professional impact—these promising trends require validation in larger, longitudinal studies. Future research with 150–200 participants per cohort and extended intervention periods is expected to confirm and reinforce these correlations, producing statistically robust evidence of the pedagogical benefits of AIGC-A/R/Tographic instruction.

VI. DISCUSSION

A. How AI-Generated Art Facilitates Cross-Cultural Dissemination of ICH

This study demonstrates that AIGC goes beyond being a technical facilitator, serving as a key bridge for cultural adaptation in the cross-cultural dissemination of ICH motifs [11]. AI expedites the digital transformation of ICH motifs, enabling quick form reconstruction and style transfer. As an AI Translation Agent, it establishes visual connections between cultural symbols, thereby enhancing the adaptability of ICH patterns to target markets [29]. Furthermore, AI expands students’ understanding of ICH by facilitating the exploration of symbolic systems through experimentation, thus advancing contemporary expressions of traditional heritage [30].

B. Expanding the A/R/Tography Framework to AI+ART

A/R/Tography emphasizes the integration of art (A), research (R), and teaching (T), and the inclusion of AI introduces new possibilities within this framework. In the artistic creation (A) dimension, AI provides efficient visual production tools that allow students to quickly construct and reconfigure ICH motifs, although manual refinement remains necessary to maintain cultural accuracy. In the research (R) dimension, AI systematized the symbolic deconstruction, style classification, and contextual adaptation of ICH motifs, expanding technological approaches in artistic research [20]. In the teaching (T) dimension, AI enhances the dynamic learning process of creation, feedback, and refinement, thus enriching the practical application of A/R/Tography. This study emphasizes that AI is not only an assistive tool in art education but also a central driving force for the deeper integration of creation, research, and teaching. Within the A/R/Tography framework, AI facilitates interactive learning by guiding students through symbolic interpretation, cross-cultural adaptation, and market validation, thus transforming cultural cognition into contemporary design applications.

C. Limitations of AI-Generated Art in Cultural and Creative Contexts

Despite AI’s significant potential in ICH education and

dissemination, several limitations remain. First, cultural accuracy remains a challenge, as AI can replicate traditional motif visual styles but often lacks a comprehensive understanding of the cultural context, leading to inconsistencies in historical patterns [19]. Additionally, balancing automation with creativity is difficult because overreliance on AI-generated designs may hinder creative development in students. Furthermore, the cross-cultural adaptability of AI-generated ICH motifs requires refinement; for instance, European markets prefer geometric designs, whereas Asian markets favor symbolic narratives, necessitating manual adjustments to ensure the designs align with specific cultural preferences [24].

D. Future Research Directions

Although this study confirmed the effectiveness of AIGC in ICH education and cross-cultural dissemination, several areas require further exploration. Future research should focus on enhancing AI’s adaptability and creativity in art education and cultural transmission, particularly through the development of cultural intelligence learning, symbolic semantic optimization, and personalized AI generation. Currently, AI mainly relies on visual feature recognition for motif generation, but it lacks a deep contextual understanding [7]. Future studies could incorporate cross-cultural corpus training and knowledge graphs to improve AI’s ability to interpret symbolic meanings and adapt contextually, enabling it to move beyond mere visual replication and more accurately convey cultural significance [30].

Another significant challenge is symbolic semantic optimization, as current AIGC primarily uses style transfer techniques that often fail to capture the deeper cultural narratives behind ICH motifs [1]. Advancements in symbolic deep learning could improve AI’s ability to analyze and understand ICH motifs more effectively, ensuring that the generated designs accurately reflect their historical symbolism rather than merely serving as decorative elements.

Furthermore, future research should explore the teaching potential of AI in human-machine co-creation, facilitating the transition from technical assistance to tools that stimulate creativity. Currently, AI is mainly used for pattern generation but has yet to fully realize its potential as a guiding force in the learning process [31]. When AI functions as a dynamic material library, it can guide students in deconstructing symbols, reorganizing cultural elements, and redesigning them, thus enhancing their cross-cultural understanding and visual expression. By adopting the “AI generation + student secondary creation” model, this approach can improve creative efficiency while stimulating students’ artistic thinking and cultural identity, ultimately promoting the innovative dissemination of ICH patterns in educational settings [25].

VII. CONCLUSION

This study explores the application of AIGC in ICH art education by developing a cross-cultural digital illustration curriculum that integrates A/R/Tography theory with Generative AI (AIGC). The findings demonstrate that AIGC enhances the efficiency of digital ICH motif creation, while the AI Translation Agent optimizes the cross-contextual adaptation of cultural symbols. This process allows students

to gain a deeper understanding of the symbolic meanings of ICH motifs and engage in creative reinterpretation based on AI-generated designs [15]. The teaching experiment highlights AI's potential in art education, showing its role not only as a creative tool but also as a digital pedagogical model that fosters cross-cultural communication.

The primary contribution of this study lies in integrating A/R/Tography theory with AI-generated art and introducing an innovative instructional framework centered on cultural symbol extraction, AI-generated optimization, artistic creation, and market adaptation. Initially, through semiotic analysis and style classification, AI learns and simulates decorative languages from different cultures to achieve cross-cultural deconstruction and the reconstruction of ICH motifs [31]. Subsequently, students refine AI-generated patterns using Photoshop and Procreate to enhance cultural expressiveness and market adaptability. These results imply that AI-generated ICH motifs have broad application potential in cultural product design, UI design, and other creative industries, ultimately achieving greater visual recognition in global markets [13].

Despite validating the value of AIGC in ICH education and cross-cultural dissemination, this study highlights the need for further improvements in AI's cultural comprehension and adaptability. Future research should focus on culturally intelligent learning, symbol semantic optimization, and personalized AI generation to enhance AI's ability to accurately interpret visual symbols, update symbol decoding for greater authenticity and adaptability, and dynamically adjust ICH motifs for different cultural markets to improve global dissemination.

In summary, this study provides both theoretical and practical insights into the digital transmission of ICH, AI-powered art education, and cross-cultural creative industries. As an intelligent creative assistant, AIGC not only enhances the digital representation of ICH and explores new pathways for its global dissemination. With continuous advancements in AI, the synergy between AI and ART is expected to play an increasingly significant role in art education and the cultural industries.

CONFLICT OF INTEREST

The authors declare no conflict of interest.

AUTHOR CONTRIBUTIONS

Xi Zhu was responsible for the research design, data collection, preliminary analysis, and manuscript drafting. Bili Duan conducted the literature review, provided key theoretical support, and performed a survey. Xiaojing Cao designed the teaching plan and course structure and implemented the teaching content. Yue Tan supervised the data collection process, analyzed experimental data, built models, and provided final review support. All authors discussed the research findings and approved the final manuscript.

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