

I Introduction

Motivations

- Artificial Intelligence (AI) advancements have impacted various sectors, including **Cultural Heritage (CH)** and **Creative Industries (CI)**
- **Neural Rendering (NR)** techniques, such as **Neural Radiance Fields (NeRFs)** and **3D Gaussian Splatting (3DGS)**, improve digitization of 3D objects from 2D images
- **Ethical concerns** remain largely **unexplored**, particularly around the authenticity, intellectual property, and the data-driven decision implications of NR-generated content.

Key Objectives

- Identify **ethical pitfalls** of NR paradigms in CCI
- Design an **ethical framework** based on global guidelines
- Provide multidisciplinary **guidelines** for developing NR solutions

Acknowledgements

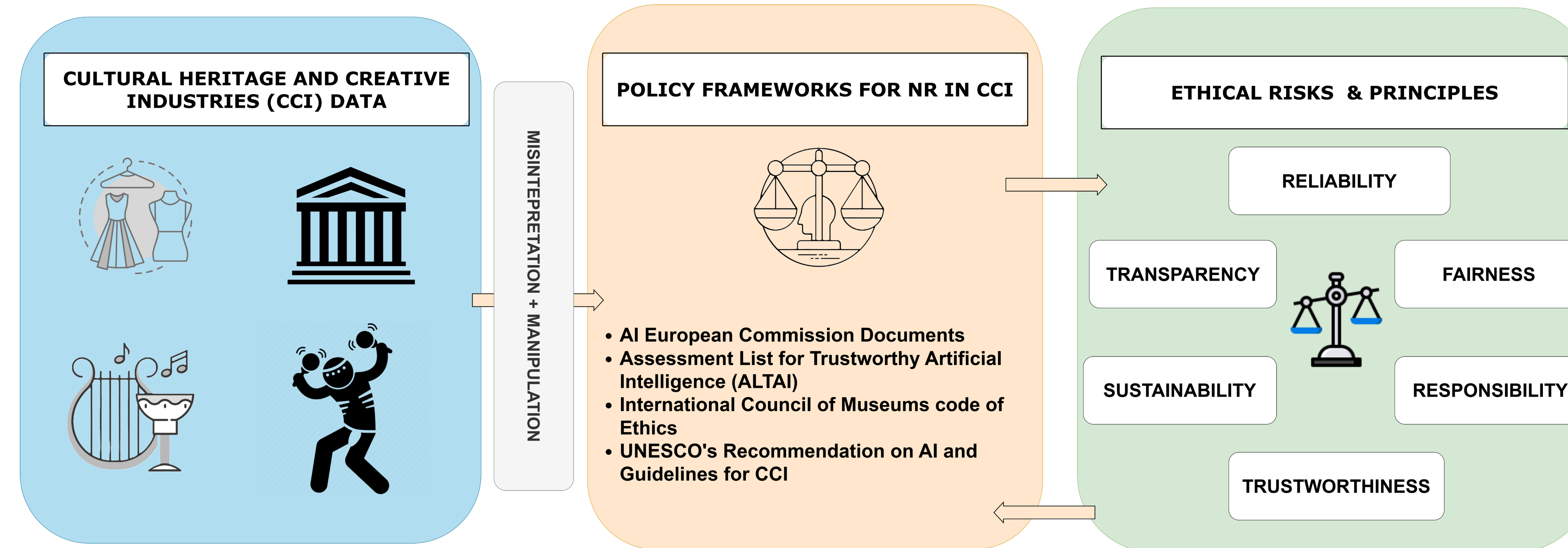


CTESQUARE
#CasaTecnologieEmergentiPesaro

CTE Square Pesaro
CUP D74J22000930008
FSC MISE 2014-2020

II Methodology

Framework

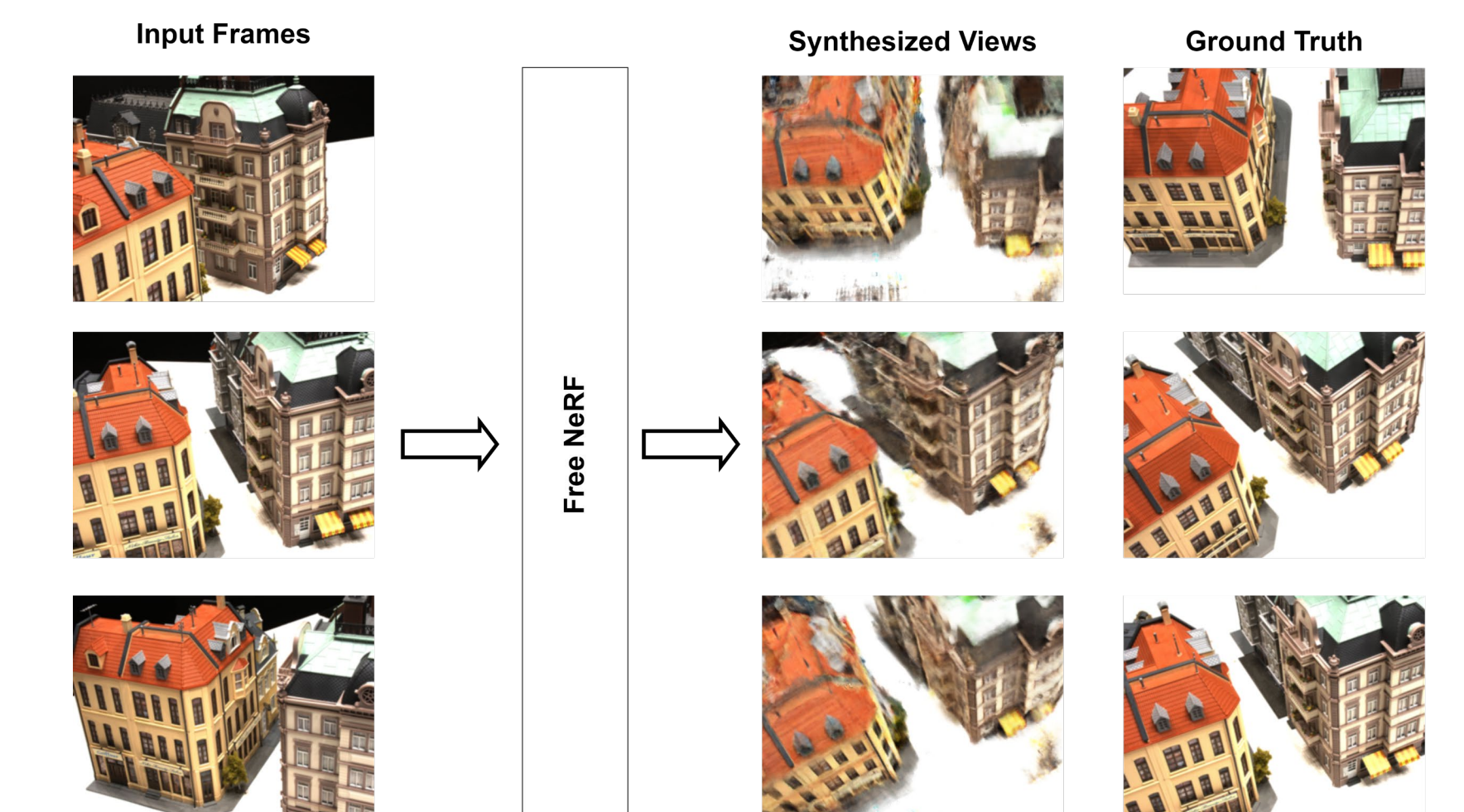


Ethical Principles in NR

- **Transparency and Explainability**: Clear communication of processes, algorithms, and outcomes
- **Reliability**: Accuracy and consistency of NR reconstructions
- **Trustworthiness**: Stability and generalization in various environments
- **Sustainability**: Minimizing environmental impact
- **Fairness**: Ensuring unbiased and fair results
- **Responsibility**: Ethical data ownership and authenticity

III Results

Ethical Principle	Challenges	Technical Risks	Possible Solutions
Transparency and Explainability	Understanding complex AI models and validate data collection process	<ul style="list-style-type: none"> • Lack of interpretability • Missing description of data collection steps • Lack of controllability for erroneous reconstructions 	<ul style="list-style-type: none"> • Provide clear and detailed documentation of data collection processes • Use visual and interpretative methods to make model decisions understandable • Maintain open communication about processes and outcomes to stakeholders.
Reliability	Ensuring accuracy of reconstructions	<ul style="list-style-type: none"> • Inconsistent outputs due to few or one-shot • Hard camera estimation due to data scarcity • Novel view synthesis and geometrical outputs with low veridicity • Bias of pre-trained NR methods 	<ul style="list-style-type: none"> • Establish rigorous testing protocols to ensure accuracy and consistency • Collect comprehensive data, ideally 50-150 images per object • Conduct bias analysis on training data and pre-trained models.
Trustworthiness	Demonstrating stability and generalization in different (social) environments	<ul style="list-style-type: none"> • Lack of visual generalization • Inconsistent Geometrical Representation • Missing social considerations into the system's functionality 	<ul style="list-style-type: none"> • Develop frameworks to demonstrate model reliability and stability • Integrate social considerations and collaborate with domain experts.
Sustainability	Minimizing environmental impact	<ul style="list-style-type: none"> • High computational demand • Energy cost to create and maintain a capture setting 	<ul style="list-style-type: none"> • Optimize model architectures and training processes to reduce energy consumption • Use renewable energy and energy-efficient hardware • Implement protocols to minimize the number of cameras and optimize GPU usage.
Fairness	Unbiased and fair results	<ul style="list-style-type: none"> • Biased NR prior knowledge • Artifacts caused by NR paradigms which exploit regularization, synthetic generation or ignore high-frequency details 	<ul style="list-style-type: none"> • Use auxiliary networks to detect and correct biases • Ensure datasets are diverse in culture, periods, and regions • Enhance model architectures and optimization strategies
Responsibility	Ethical data ownership and authenticity	<ul style="list-style-type: none"> • Misuse of generated data • Accountability for unfaithful generation • Intellectual property 	<ul style="list-style-type: none"> • Follow ethical data ownership and secure handling protocol • Implement validation protocols to ensure data authenticity • Adhere to legal and intellectual property frameworks and protect data integrity.



Reliability on few-shot models: FreeNeRF network trained on 3 images from the DTU dataset with the same setting provided by the original authors and 3 synthesized novel views compared against their ground truths