

## MACHINE LEARNING FOR THE DOCUMENTATION, PREDICTION, AND AUGMENTATION OF HERITAGE STRUCTURE DATA

Satwant Rihal\*, Hisham Assal

Cal Poly State University, San Luis Obispo, California, USA – (srihal, hhassal)@calpoly.edu

**KEY WORDS:** Heritage structures, Survey and Documentation, Heritage Data, Machine Learning, Natural Hazards, Ontology

### ABSTRACT:

The paper presents an effort to develop learning models based on the massive amounts of data that has been accumulated over the past decades during the process of digital documentation of heritage structures around the globe especially those in disaster zones. The development of an ontology is proposed that describes heritage buildings, their sites, and major hazard events that may cause damage to them. This ontology can serve as a repository for documenting heritage structures and provide highly structured data for developing machine learning systems that can identify patterns of damage from recorded image data. For heritage structures in seismic zones, the first step in ontology development is analyzing available earthquake information about the event and the damage information. The resulting model will create links between information items, for example relating the extent of the damage of an element to the earthquake magnitude and its distance from the epicenter. The ontology may also include collected images from previous earthquake events, with links to the objects in each image. Special tools will focus on selecting sub-models to be included in a machine learning model. For example, if the learning objective is to identify the damage and its extent from an image, then the rules will select the features in the model that relate to structural damage and identify each type of damage. It is hoped that this work will help develop learning systems that speed up processing of large volumes of image damage data collected from heritage sites.

### 1. INTRODUCTION

During the past thirty years the world has experienced unprecedented levels of devastation, loss of innocent lives and staggering losses caused by catastrophic earthquakes as presented in Table 1.

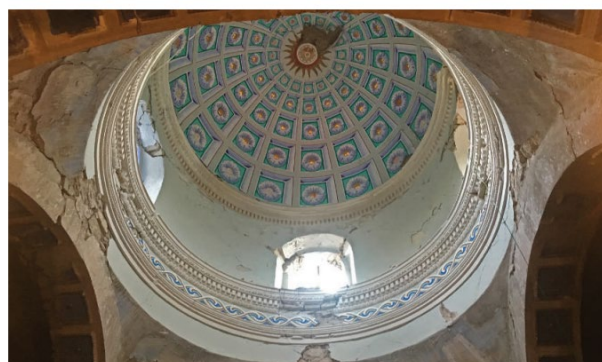
Location	year	Magnitude	Fatalities	Economic Losses B\$
Kobe, Japan	1995	6.9	> 6000	197
Afghanistan	1998	6.6	4000	
Izmit, Turkey	1999	7.6	20000	19
Gujarat, India	2001	7.6	20000	5
Indonesia	2004	9.1	230,000	15
Sichuan, China	2008	7.9	87,500	148
Haiti	2010	7.0	316,000	14
Chile	2010	8.8	521	30
Tohoku, Japan	2011	9.0	18,400	360
Christchurch, New Zealand	2010	7.1		40
Emilia-Romagna Italy	2012	6.1	27	16
Nepal	2015	7.8	8,800	10
Central Italy	2016	6.2	247	0.208 Euros (insured)
Central Mexico	2017	7.1	>370	4
Turkey/Syria	2023	7.8	57,658	104

**Table 1.** Major earthquakes over the last 30 years

The resulting aftermath of human toll of hundreds of thousands of lives lost, and economic losses amounting to billions of dollars seriously affected countries' economies in different

sectors, such as housing, infrastructure, health, education, industry, cultural heritage and tourism.

With the recent advancements in digital technologies and tools available for the documentation of heritage structures, (e.g. satellites, UAV, TLS, drones, GPS, and GIS), massive volumes of data have been collected and accumulated over the recent decades during the survey and documentation process of field reconnaissance and observed damage following devastating earthquakes, (Hutchinson, 2017), (Hadick, 2022). The data covers a wide range of heritage structures in many regions around the world. The sheer volume of data makes it difficult to comprehend the extent of coverage and utilize it in a useful way.



**Figure 1.** Typical damage of domes – Historical Churches September 2017 Puebla, Mexico earthquake

\* Corresponding author



**Figure 2.** Typical damage of facade and bell towers – Historical Churches September 2017 Puebla, Mexico

The earthquake hazard assessment and seismic retrofit of buildings is a complex problem on both the architectural and engineering levels. It involves many related issues and variables. These issues cannot be considered in isolation. Some of the attributes may be partially defined and some of the required information may be incomplete or missing. Decisions have to be made on the basis of such uncertainties.

There is a need and an opportunity to develop learning models based on the massive volumes of data acquired through detailed digital documentation of heritage structures especially in disaster zones around the globe. The paper presents an ontology-based environment for critical assessment of the available digital data from the heritage documentation process. Handling such a large volume of data requires a common representational model, such as an ontology, to relate the data elements and identify overarching concepts that are present within the data. The common ontology facilitates the development of systems that provide analysis of the data and extraction of useful patterns, such as machine learning systems. The results will provide an intelligent framework for systematically deriving and updating lessons from observed earthquake damage data, and from databases of heritage documentation processes.



**Figure 3.** Typical damage of facade and bell towers – Historical Churches September 2017 Puebla, Mexico earthquake

The intelligent framework will manage the complexity of analysis, in collaboration with the human user, by creating information models for buildings, over which deep learning systems can operate to analyze the building conditions and provide feedback to users. The internal model is based on a domain ontology to represent building elements and the rich and complex relationships that exist among them. By embedding knowledge into the building analysis environment, we provide context for the analysis tools to produce more meaningful results and recommendations to the user. The information model represents the building under analysis and the deep learning systems examine different aspects of the building and its affected elements, such as significant architectural features and elements

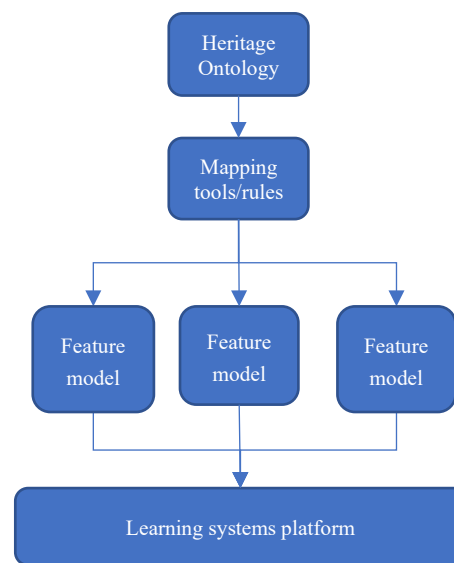
(e.g. domes, bell towers, facades, ornamentations, etc.) and assess their cultural value and their need for restorations.

## 2. OBJECTIVES

This paper proposes to develop an ontology to represent all aspects of heritage structures and their values, and use the ontology to generate machine learning (ML) models for specific types of analyses as needed. The ontology allows for the definition of rules that specify required parts of the general model to be included in each machine learning model. There are three components to this proposed development: 1) the ontology, 2) mapping tools and rules, and 3) the learning system platform. Figure 4 shows the general architecture of the proposed system.

## 3. PREVIOUS WORK

Over the past ten years the authors have been actively pursuing the development of intelligent knowledge-based systems for extraction of lessons from the massive body of knowledge that has been accumulated from the observations of the performance of existing buildings in general, and cultural heritage structures in particular, during catastrophic earthquakes around the world over the past sixty years and beyond (Rihal & Assal, 2012).



**Figure 4.** Ontology-based machine learning system Architecture

In the early work an intelligent framework was developed to support the critical assessment of the available body of knowledge, and the development of intelligent design agents, that serve as analysis tools in different areas of the seismic design process (Rihal & Assal, 2016).

This work was followed by the development of an ontology-based environment for earthquake analysis and design, involving identification and integration of data sources, developing a unified model of building information and risk information, intelligent analysis tools and a system for providing access to the unified model for data consumers (Rihal et al. 2020).

An effort to better understand the types of data needed for machine learning systems, lead to further research into the information needs for a deep learning system for the assessment and restoration of earthen heritage structures (Rihal & Assal, 2022).

Recent comprehensive work on automated processing of earthquake damage image data has been presented by (Yeum, et. al. 2016) (Yeum, et. al. 2017) and (Yeum, et. al. 2018). It is based on machine learning and computer vision, to classify and organize large volumes of visual data, in support of post-earthquake reconnaissance.

An annotated damage image database to support AI-assisted Hurricane Impact analysis has been presented by (Ou, et. Al. 2021).

A deep learning algorithm was implemented by (Patterson et. al. 2018). It supports automated image classification of seismic damage to built infrastructure, and identification of multiple damage types and associated structural members in a single image

The recent works cited above deal with seismic damage data for structures in general and do not include damage data for heritage and historical buildings.

#### 4. HERITAGE DATA

Heritage structures represent a great value for its society and serves as documentation for history and culture. The damage caused by earthquake events to heritage structures may lead to loss of this documentation and the cultural value associated with it. It's imperative that such structures be preserved, documented, and restored in the event of damage. In addition to structural details of building elements, heritage data include dates, historic significance, association with prevailing culture at the time of construction, and other important aspects. The diversity of data formats (e.g. pictures, videos, textual descriptions, etc.) and the organizations performing the documentation work require that there be a standard for describing elements of structures, so that collaborative work can take place and restorative work has a point of reference of how the structure looked before the damage.

##### 4.1 Advanced technologies for digital documentation of heritage structures

Over the past fifty years great progress has taken place in the application of advanced technologies e.g. UAV, drones, terrestrial photogrammetry, aerial photogrammetry, terrestrial laser scanning (TLS), airborne laser scanning (ALS), LiDAR, cameras, GIS etc., as evidenced by (Santana 2022); (Jigyasu et. al 2022); (Kallas, 2021); (ADRC 2016); (Croce, et. al., 2021); (Servin, 2010); (ICCROM, 2022); (Bartlett et. al., 2020); (Altan, et. al., 2001); (Chatzistamatis, et. al, 2018); (Stepinac, et.al , 2020); (Rouhani, et. al., 2020); (Monical, 2020); (Costamagna, E., et. al., 2019); (Kaartinen, E. et. al, 2022); (Shrestha, et. al. 2017).

Excellent contributions and leadership to the field of information technologies and 3D digital documentation of cultural heritage before and after disasters and pandemics around the world, have been presented by (Santana, 2021)

The emergency 3D documentation and photogrammetry survey of the cultural heritage buildings damaged by the 2020 explosions in Beirut, identification of buildings at high risk of collapse, and development of conservation plans for the heritage buildings has been presented by (Kallas, 2021).

A very exciting project on monitoring endangered archaeology in the Middle East and North Africa (EAMENA) has been presented by (Rouhani, 2020). This project involved creation of a digital record of archaeological sites into the EAMENA

database to improve the inventory, management and preservation of archaeological sites. The sources of information for this project are remote sensing; published reports, existing databases and archives; and data collected from field surveys and field assessment.

In general, the digital workflow can be represented as shown in the Figure 5.

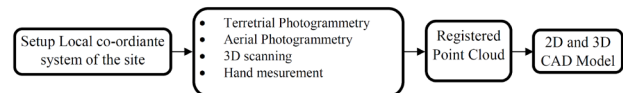


Figure 5. Digital workflow used for the acquisition of data (Source: Shrestha et. al. 2017)

A great deal of work has been carried out in recent years on the reconnaissance and 3D documentation of heritage structures damaged by earthquakes and other disasters as outlined in the references cited above.

##### 4.2 Digital Documentation of Heritage Structures damaged during earthquakes

The most interesting type of documentation of heritage structures is the digital 3D laser scanning and photogrammetry.

###### 4.2.1 Metropolitan Cathedral, Mexico City - 3D Digital Documentation – CyArk:

One of the important sources of heritage 3D data is CyArk as presented by (Hadick, 2022). The 3D digital documentation methodology used by CyArk, for documentation of the Metropolitan Cathedral in Mexico city which suffered damage during the 2017 Puebla, Mexico earthquakes, consists of LiDAR (Light Detection and Ranging), LiDAR Registration, leading to Registered Point Cloud; Terrestrial Photogrammetry, Image Alignment and 3D Reconstruction; Aerial Photogrammetry using a DJI Drone; Aerial Image Alignment and 3D Reconstruction; LiDAR and Photogrammetry Synthesis and Photo-textured 3D Model; Architectural Drawings Extraction  
[www.youtube.com/watch?v=UeDMA-Bk\\_LM](http://www.youtube.com/watch?v=UeDMA-Bk_LM)  
[www.youtube.com/watch?v=7kILcqfAIMU](http://www.youtube.com/watch?v=7kILcqfAIMU)  
CyArk - Digitizing the Metropolitan Cathedral in Mexico City

###### 4.2.2 Changu Narayan Temple – Kathmandu Valley – 3D Documentation – Skydio 3D Scan:

A second possible source of heritage 3D data is Skydio ([www.skydio.com/blog/digital-reconstruction-nepal-changu-narayan-temple](http://www.skydio.com/blog/digital-reconstruction-nepal-changu-narayan-temple)). A magnitude 7.8 catastrophic earthquake struck Nepal in April 2015, resulting in massive loss of life resulting in almost 9000 deaths and thousands who were injured. About 80% of the heritage structures and temples in Bhaktapur and Kathmandu Durbar Square were destroyed during this 2015 Nepal earthquake. A 3D digital documentation of the historic Changu Narayan Temple in the Kathmandu Valley, dating back to the 5th. Century A.D was carried out by Skydio 3D Scan. After the flights the data was processed in Bentley's Context Capture software to create the Changu Narayan Temple 3D model, which establishes a baseline to compare future scans after future disaster events.

**4.2.3 2017 Mexico earthquake observed damage data – UNAM Database:** Two catastrophic earthquakes struck south and central Mexico, in September 2017, the first one on September 7 (Mw =8.2) and the second on September 19 (Mw = 7.1). It was reported that approximately 2340 heritage buildings, amounting to about 6% of all the heritage buildings in the areas affected by the two devastating earthquakes were damaged (Peña & Chávez, 2021). After these devastating earthquakes a survey and reconnaissance of the damage suffered by historical temples was undertaken by a team of the Institute of Engineering at UNAM. The results of the survey and reconnaissance of observed performance and damage of historical temples in Mexico have been presented by (Chavez et. al. 2021), (Pena et. al. 2021), (Pena et. al. 2016), (Pena et. al. 2020), (Pena et. al. 2021), (Preciado et. al. 2022), (Preciado, et. al. 2020), (Weiser et. al., 2018). A total of 58 temples were visited: 22 in Oaxaca, around the Mixteca Alta and Tehuantepec regions; 11 inside the Mixteca region, Puebla; 15 in Morelos, and 10 in Mexico City. Some of these temples are included in the UNESCO World Heritage List of UNESCO.

A preliminary seismic damage and vulnerability assessment of Mexican churches after the September 2017 earthquakes was also presented by (Diaz, et. al., 2019).

A database was developed by the team at UNAM to manage the large amount of information collected during the damage survey and reconnaissance program. The database was developed by using the Microsoft Access software. The database was organized in such a way that it can be searched in three different ways: state or location, type of roof system and shape of the nave. Once the search is selected, the list of churches that match the query criteria is displayed. Clicking on the name of the selected church, the record of each church is accessed. This record is divided in five sections. In the first section, the name of the church and its location is displayed. The second section is about the photos of the building available in the database. For each record, there are photos of the state of the structure before the 2017 earthquakes, of the damage observed by these earthquakes and of the rehabilitation project (in case there is information about it). In the third section, complementary files, as documents, videos or draws, can be found. The fourth section give information about the characteristics of the church, as well as the damages reported for each microelement: façade, nave, transept, apse and dome. Finally, the record ends with a brief overview of the damage observed in the building, as well as the reinforcing actions implemented (in case there is information about it). Further details of the UNAM database are presented elsewhere (Rihal, et. al. 2023).

**4.2.4 Damage Survey on Churches – Observed Damage Database of past earthquakes (Da.D.O.):** An IT platform named "Da.D.O." (Database of Observed Damage) to collect, catalogue and compare data on structural characteristics and the observed damage of buildings during earthquakes in Italy from the 1976 Friuli earthquake onwards has been presented (Di Meo et. al., 2023). Details of the functioning of Da.D.O. section on churches, a detailed description of the databases and the survey forms used in post-earthquake inspections of churches, from which the databases were derived are presented therein, including some results about the most common typologies, structural characteristics and the damage mechanisms that were activated during the seismic events.

An analysis of the observed damage to thirty-six masonry churches during the 2016 Central Italy earthquake has been presented by (Ferracuti et. al. 2022). The analysis highlighted the most frequent damage mechanisms and the most vulnerable macroelements, including development of a damage index based on the observed damage and the macroelements present in the

surveyed churches.

### 4.3 Information Sources

Many databases exist with organizations that track and record earthquake information in different ways, ranging from general earthquake event data, such as location, epicenter, magnitude, etc. (e.g. USGS) to damage data to structures and more specifically damage to cultural heritage buildings (e.g. World Monument Fund, ICOMOS, CyArk, UNAM, INAH). Figure 6 lists some of the more prominent data sources around the world, which provide access to researchers and earthquake experts.

One of the first disaster and failure studies data repository was the NIST database for the 2010 Chile earthquake as presented by (Catlin and Pujol 2015).

A general-purpose natural hazard image database of different hazards e.g. earthquake, tsunami, volcano, landslide, geology is hosted by (NOAA, 2023).

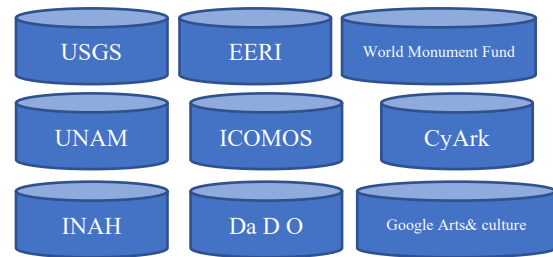


Figure 6. Earthquake data sources

## 5. ONTOLOGY DEVELOPMENT

An ontology is a description of a field of study in terms of objects and their relationships with some rules of inference, which describe implicit relationships. It may be developed in a hierarchy or in a graph-like structure. A well-developed ontology is the basis of intelligent applications and tools in addition to supporting the creating of machine learning models for many purposes. We propose the development of an ontology that describes heritage buildings, their sites, and major hazard events that may cause damage to them. This ontology can serve as a repository for documenting heritage structures and provide highly structured data for developing machine learning systems that can identify patterns of damage from images, predict the type of damage that may occur in the case of a given event, such as an earthquake, and support efforts of renovation, restoration, or conservation.

The ontology has four main areas: buildings, hazard events, sites and regulations. It represents all the available information about structures in earthquake zones, related to earthquake events and the damage that may have been caused by it. The regulations component of the ontology reflects the cultural value of structures through the requirements of preservation and repair efforts that must be performed on damaged structures. The cultural value can be separated into its own section of the ontology to reflect values that may be designated by non-governmental organizations.

The four areas of the ontology will be connected to describe statements such as: "Structure 'A' in location 'B' sustained damage of elements [a, b, c] because of event 'Earthquake E'"

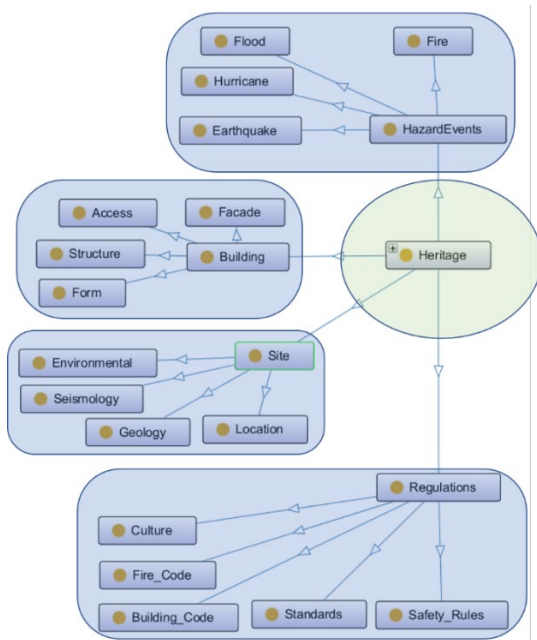


Figure 7. Sample Ontology for Heritage Structures

Rules can be declared in the ontology to define conditions under which class objects can have a relationship to other objects in the same class or in other classes. These rules can be used to select a subset of the ontology to be included in a machine learning model. For example, a ML system that aims to learn patterns of structural damage in buildings may include attributes of the buildings class which describe structural elements and elements from the hazard events class that describe the forces generated by each event and the stresses that impact the structural elements.

## 6. METHODOLOGY

The ontology development starts with an analysis of available earthquake information, both event information and damage information. This analysis should lead to the design of a model that encompasses aspects of that are represented in the information. The model will create links between information items, such as relating the extent of damage of an element to the earthquake magnitude and distance from the epicenter.

Tools will be developed to work with the model to allow ontology navigation. Some tools will allow exploring the objects and their relationships in the model. Other tools will allow the definition of rules or constraints to select specific objects of interest to examine. The ontology may also include collected images from previous events, with links to the objects in each image. This part may facilitate the creation of machine learning models later.

Special tools will focus on selecting sub-models to be included in a machine learning model. For example, if the learning objective is to identify the damage and its extent from a picture, then the rules will select the features in the model that relate to structural damage and identify each type of damage.

Figure 5 shows the general steps for creating a machine learning system. The proposed ontology and its tools will assist in the steps of 'feature extraction' and 'model development'

## 7. CREATION OF MODELS FOR ML SYSTEMS

Machine learning systems require large volumes of data prepared and organized into models that describe the system's interest (i.e., what the system is trying to learn) and the significant attributes of the objects involved. Given the variety of learning objectives and the number of machine learning systems that can be built with the collected data, it would be hard to develop each model by hand from scratch. The highly structured nature of the ontology allows us to build tools that would extract a desired model out of the relevant portion of the ontology. We propose building tools that would take in model requirements, generate an appropriate model from the ontology, and populate it with data as appropriate.

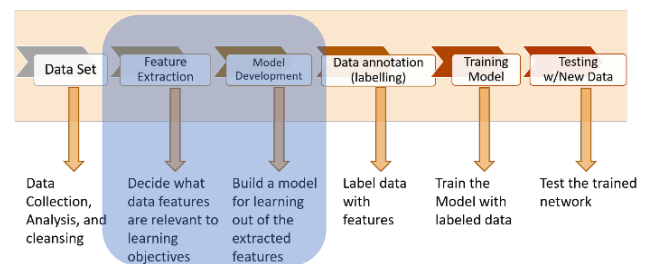


Figure 8. Steps of creating a machine learning system

## 8. OUTCOMES AND BENEFITS

This work will provide an overall ontology for the earthquake (and general hazard) events, which will relate the events to the damage they cause to structures. It will also aid in developing learning systems that speed up the processing of large volumes of images collected from damaged sites to make more effective use of the massive amounts of recorded damage data for decision making for conservation of cultural heritage buildings in hazard zones around the globe.

## REFERENCES

- Altan, Orhan, et. al., "Photogrammetry and geographical information systems for quick assessment, documentation and analysis of earthquakes", ISPRS Journal of Photogrammetry and Remote Sensing, March 2001
- Arakida, and Nakao 2016 Kumamoto Earthquake Survey Report, [www.adrc.asia/publications/201604\\_KumamotoEQ/ADRC\\_2016KumamotoEQ\\_Report\\_1.pdf](http://www.adrc.asia/publications/201604_KumamotoEQ/ADRC_2016KumamotoEQ_Report_1.pdf), Asian Disaster Reduction Center (ADRC)
- Bartlett, F.M., Turkel, W.J., "Digital Analysis of Historic Bridge Images", Proceedings 26th. International Conference on Cultural Heritage and New Technologies, CHNT, 2020
- Catlin, A.C., & Pujol, S., "NIST Disaster and Failure Studies Data Repository: The Chile Earthquake Database-Ground Motion and Building Performance Data from the 2010 Chile Earthquake – User Manual", National Institute of Standards and Technology (NIST), December 2015
- Chatzistamatis, S., et. Al., "Fusion of TLS and UAV Photogrammetry Data for Post-Earthquake 3D Modeling of a Cultural Heritage Church", ISPRS, Vol. XLII-3/W4, 2018, Geoinformation for Disaster Management, March 2018, Istanbul, Turkey

Chavez, Marcos; Peña, Fernando; Garcia, Natalia, & Duran, Daniel, "Damage Patterns in Historical Temples of Puebla, Morelos and Oaxaca after September 2017 Mexico Earthquakes", 12th. International Conference on Structural Analysis of Historical Constructions, SAHC2020, Barcelona, Spain Oct. 2021. P. Roca et. al. (editors)

Costamagna, E. et. al., "Advanced non-destructive techniques for the diagnosis of historic buildings: The Loka-Hteik-Pan temple in Bagan", *Journal of Cultural Heritage* (2019), <https://doi.org/10.1016/j.culher.2019.09.006>

Croce, Valeria et. Al., "Seismic Damage evaluation and decisions on interventions supported by UAV-based surveys", PROHITECH 2020, 4th. International Conference on Protection of Historical Constructions, Oct. 2021, Athens, Greece. <https://hal.science/hal-03406144/document>

Di Meo, Antonella; Faravelli, Marta; Pascale, Venanzio, Borzi, Barbara; Calderini, Chiara; Sisti, Romina; Speranza, Elena; Bocchi, Flavio. Damage survey on churches: A new Observed Damage Database of past Italian earthquakes (Da.D.O.), *International Journal of Disaster Risk Reduction*, Volume 87, 2023.

Diaz, Daniela, Baquedano, Pilar, D'Amato, Michele, & Laterza, Michelangelo, "Preliminary Seismic Damage Assessment of Mexican Churches after September 2017 Earthquakes", *International Journal of Architectural Heritage*, (2019)

Ferracuti, Barbara; Imperatore, Stefania; Zucconi, Maria & Colonna, Silvia, "Damage to Churches after the 2016 Central Italy Seismic Sequence", *Geosciences* 12, 122 (2022)

Hadick, K., 2022 Open Heritage 3D, ([www.CyArk.org](http://www.CyArk.org)), CyArk Inc., San Francisco, California.

Hutchinson, Tara, Franke, Kevin, & Mayoral Villa, Juan Manuel, "Geotechnical Engineering Reconnaissance of the 19 September 2017 Mw 7.1 Puebla-Mexico City Earthquake", GEER Report Version 2.0, 2020 ([www.geerassociation.org](http://www.geerassociation.org))

Jigyasu, Rohit and Ikawa, Hirofumi, (moderators), "Applying Digital Documentation – Sustainable Heritage Preservation and Management" <https://www.iccrom.org/events/applying-digital-documentation-sustainable-heritage-preservation-and-management>, ICCROM October 2022

Jigyasu, Rohit, et. Al., "3D Laser scanning documentation for informing the post-earthquake recovery of heritage settlements – a Practical Guide", ICCROM [www.irep.ntu.ac.uk/id/eprint/47306/1/10.17631-rd-2022-0005-dprac\\_v2.pdf](http://www.irep.ntu.ac.uk/id/eprint/47306/1/10.17631-rd-2022-0005-dprac_v2.pdf)

Kaartinen, E. et. al., "LiDAR-Based Structural Health Monitoring: Applications in Civil Infrastructure Systems", *Sensors* 2022, Vol. 22, 4610 <https://doi.org/10.3390/s22124610>

Kallas, Joe., "Emergency Documentation Mission of Historic Beirut Post-Explosion", *Globinar, Our World Heritage*, Amsterdam, May 2021

Monical, J. (2020), "Building Surveys after Earthquakes," <https://datacenterhub.org/deedsdv/publications/view/137>

NOAA, "Natural Hazards Image Database", <https://ngdc.noaa.gov/hazardimages/>#, National Oceanic and Atmospheric Administration, last accessed 4-29-2023.

Ou, H. N.; Ro, S. H.; Gong, J. and Zhu, Z., "Building an Annotated Damage Image Database to Support AI-Assisted Hurricane Impact Analysis," 2021 IEEE International Conference on Imaging Systems and Techniques (IST), Kaohsiung, Taiwan, 2021, pp. 1-6, doi: 10.1109/IST50367.2021.9651432.

Patterson B.; Leone, G.; Pantoja, M.; and Behrouzi, A., "Deep Learning for Automated Image Classification of Seismic Damage to Built Infrastructure", 11th. U.S. National Conference on Earthquake Engineering, June 2018, Los Angeles, California

Peña, Fernando and Chavez, Marcos, "Inadequate Cases of Intervention in Architectural Heritage Buildings in Mexico after the September 2017 Earthquakes", 12th. International Conference on Structural Analysis of Historical Constructions, SAHC2020, Barcelona, Spain Oct. 2021. P. Roca et. al. (editors)

Peña, Fernando; Chavez, Marcos, & García, Natalia, "Mexican colonial churches: structural assessment and seismic behavior", Chapter 12, in *Masonry construction in active seismic regions*, Elsevier, 2021. R. Rupakhety & D. Gautam (editors)

Preciado, Adolfo, Peña, Fernando, Fonseca, Fabiola, & Silva, Citlalli, "Damage description and schematic crack propagation in Colonial Churches and old masonry buildings by the 2017 Puebla-Morelos earthquakes (Mw =8.2 and 7.1), *Engineering Failure Analysis* 141 (2022)

Preciado, Adolfo; Santos, Juan; Silva, Citlalli; Ramirez-Gaytan, Alejandro; and Falcón, José. "Seismic damage and retrofitting identification in unreinforced masonry Churches and bell towers by the September 19, 2017 (Mw=7.1) Puebla-Morelos Earthquake, *Engineering Failure Analysis*, 118 (2020)

Rihal, Satwant and Assal, Hisham. (2012), "A Database Integration Approach to Support Earthquake Hazard Assessment and Seismic Retrofit of Buildings", *Proceedings, 15th World Conference on Earthquake Engineering*, Lisbon, Portugal

Rihal, Satwant and Assal, Hisham. (2016), "An Intelligent Framework to Support the Seismic Hazard Mitigation of Heritage Structures in New Delhi", *Proceedings, International Conference on Urban Risks (iCUR)*, Lisbon, Portugal.

Rihal, Satwant; Assal, Hisham; Badillo, Hiram and Lagunes, Maria Margarita Segarra (2020) "A Deep Learning System for The Assessment and Restoration of Heritage Structures: Case study of the 2017 Puebla, Mexico Earthquake". ICOMOS GA 2020 6-ISCs Joint Meeting: Advancing Risk Management for the Shared Future. Live Webinar, 2022.

Rihal, Satwant and Assal, Hisham. (2022). "Information Needs for A Deep Learning System for The Assessment and

Restoration of Earthen Heritage Structures: Case Study of the 2017 Puebla, Mexico, Earthquakes". Terra 2022, 13th World Congress on Earthen Architectural Heritage, Proceedings, Los Angeles, California, Getty Publications, forthcoming

Rihal, Satwant and Assal, Hisham. (2023). "Machine Learning for the Documentation, Prediction, and Augmentation of Heritage Structure Data", Proceedings, CIPA 2023 (in preparation)

Rihal, S. et. al., "Deep Learning Modelling of Earthquake Damage Data for Identification of Patterns of Damage in Heritage Structures", to be presented at the SAHC 2023 Conference, September 2023, Kyoto Japan

Rouhani, Bijan., "Monitoring Endangered Archaeology in the Middle East and North Africa", Globinar #4: Information Systems: Protecting the Past, Securing the Future February 7, 2020, Our World Heritage Foundation, Amsterdam, Netherlands

Santana, M., "Information Technologies for Preparedness and Mitigation: Prior or After Disasters and Pandemics", Globinar, Our World Heritage Foundation, Amsterdam, Netherlands May 2021

Servin, Victoria., "New Techniques for Earthquake Reconnaissance", 2010  
<https://peer.berkeley.edu/sites/default/files/2010-servin-paper-final.pdf>

Shrestha, S., et. al, "Digital Recording and Non-Destructive Investigation of Nayatapola Temple after the Gorkha Earthquake of 2015" ICOA 1687 ...need citation

Stepinac, Mislav and Gasparovic, Mataeo., "A Review of Emerging Technologies for an Assessment of Safety and Seismic Vulnerability and Damage Detection of Existing Masonry Structures", Applied Sciences, 2020 Vol. 10, DOI:10.3390/app10155060

Weiser, Deborah; Hunt, Jeffrey; Jampole, Ezra; and Gobbato, Maurizio, "M7.1 Puebla, Mexico Earthquake on September 19, 2017", EERI Earthquake Reconnaissance Team Report, Earthquake Engineering Research Institute, February 2018

Yeum, C. M. et. al., "Big visual data analytics for damage classification in civil engineering", 2016. Transforming the Future of Infrastructure through Smarter Information: Proceedings of the International Conference on Smart Infrastructure and Construction, 27–29 June 2016

Yeum, C.M., et. al., "Visual Data Classification in Post-Event Building Reconnaissance, 2017 Journal of Engineering Structures, Vol. 155, 15 January 2018, pp 16-24.

Yeum, C.M., et. al., "Automating Visual Data Processing to Support Post-Earthquake Reconnaissance", Eleventh U.S. National Conference on Earthquake Engineering, June 2018, Los Angeles, California.