

EGU24-15914, updated on 20 May 2024

<https://doi.org/10.5194/egusphere-egu24-15914>

EGU General Assembly 2024

© Author(s) 2024. This work is distributed under the Creative Commons Attribution 4.0 License.



## Machine Learning Estimator for Ground-Shaking maps

Marisol Monterrubio-Velasco<sup>1</sup>, **Rut Blanco**<sup>1</sup>, Scott Callaghan<sup>2</sup>, Cedric Bhihe<sup>1</sup>, Marta Pienkowska<sup>3</sup>, Jorge Ejarque<sup>1</sup>, and Josep de la Puente<sup>1</sup>

<sup>1</sup>CASE, Barcelona Supercomputing Center, Barcelona, Spain (marisol.monterrubio@bsc.es)

<sup>2</sup>Southern California Earthquake Center, California, USA

<sup>3</sup>ETH, Zurich, Switzerland

The Machine Learning Estimator for Ground Shaking Maps (MLESmaps) harnesses the ground shaking inference capability of Machine Learning (ML) models trained on physics-informed earthquake simulations. It infers intensity measures, such as RotD50, seconds after a significant earthquake has occurred given its magnitude and location.

Our methodology incorporates both offline and online phases in a comprehensive workflow. It begins with the generation of a synthetic training data set, progresses through the extraction of predictor characteristics, proceeds to the validation and learning stages, and yields a learned inference model.

MLESmap results can complement empirical Ground Motion Models (GMMs), in particular in data-poor areas, to assess post-earthquake hazards rapidly and accurately, potentially improving disaster response in earthquake-prone regions. Learned models incorporate physical features such as directivity, topography, or resonance at a speed comparable to that of the empirical GMMs.

In this work, we present an overview of the MLESmap methodology and its application to two distinct study areas: southern California and southern Iceland