

IP4: Harnessing Disruptive Technologies for Earthquake Resilience

Nirmal Nair and Garry McDonald





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Modelling and evaluation of

benefits and pitfalls

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Smart cities and real-time

sensing

Social Networks

Event Trees

Social Networks Causal Loon Diagr Event Trees Bayesian Networks

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Social Netwo Causal Loop Diagra Stock & Flow Diagra Event Trees

Bayesian Networks Social Networks

Causal Loop Diagram

Event Trees

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Adoption pathways for disruptive technologies

Renewable Distributed Energy

AHP Analysis for Disaster Hazard Mapping to Optimize Solar and Wind Farm Site Selection

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Constructing solar and wind power plants is a significant long-term investment the es intensive consideration. The objective is to generate optimal power while izing investment and operational expenses, all while ensuring minimal requires int environmental impact. Thus, identifying the most optimal location becomes crucial. Unfortunately, countries prone to natural disasters, including floods, landslides, earthquakes, bushfires, and storms, consistently experience damage and power loss. This study aims to explore the Analytical Hierarchy Process algorithm, which incorporates input from stakeholders and planners, to establish a comprehensive disaster mapping criterion. By identifying and prioritizing crucial parameters, we can effectively mitigate the impact of natural hazards on solar and power generation and make informed decisions for sustainable and resilient energy infrastructure. This study will perform the AHP analysis for establish flood hazard mapping, a study case of Hawkes Ba



												Nicity	-
	1.00	2.80	3.00	2.00	4.00	2.00	5.01	4,81	3.00	2.00	2.00	37.6%	1
	9.59	2.80	3.00	1.00	5.00	1.00	5.00	4.88	4.89	3.00	3.00	24.9%	1
	0.88	4.11	1.00	6.11	6.00	0.33	0.14	1.00	2.00	1.00	1.11	5.7%	,
	0.30	3.80	1.00	1.00	1.00	2.00	3.81	4.00	6.00	2.00	2.00	31.75	,
	0.25	4.30	1.21	6.30	100	0.25	0.28	0.58	0.75	1.21	8.30	2.2%	1
	9.39	4.50	3.00	8.90	4.00	100	2.01	4.01	4.00	3.00	8.90	33.2%	
	9.35	4.33	2.00	8.13	5.00	0.50	1.01	4.01	3.80	2.00	8.90	1.25	
	0.8	4.8	8.33	6.25	2.00	0.25	0.8	1.88	0.50	8.33	4.8	2.9%	,
	9.30	625	8.50	6.25	4.00	0.15	0.30	2.88	1.80	8.50	8.33	425	3
	0.30	2.00	1.00	830	4.00	0.33	0.34	1.01	2.00	1.00	8.30	7.2%	1
	0.30	6.01	1.00	810	3.00	1.00	2.00	4.00	8.00	2.00	1.00	21.6%	

Principal eigen value = 11.77









Enhanced Morlet Wavelet-Based Two-Point - M-MW-FIR FIR for Phasor Estimation - e-M-MW-FIR M-DTFGT M-class Std × 10⁻² ¥ Хен 10⁻⁴ 50 100 50 Interference frequency, Hz Interference frequency, Hz f_o, Hz These advanced instruments with accuracy, fast and robust phasor measurements in the distribution system could enable multiple applications, such as situational awareness in distribution grid, early detection of network degradation pre-



event, identification of failure hierarchy of a

renewable distributed energy system, etc.

Smart Cities

- Text may contain useful location information about earthquake impacts.
- Mapping of these impacts is important for situation response.
- To map them, we need to know where they are.
- · We use machine learning to detect whether text refers to a location that can be mapped.
- Location is more complex than just place name:
 - Tree down outside Orewa
 - Traffic congestion beyond the Bridge
 - People fleeing coastal areas across the city
 - · Wellington responded to the situation in the North
- We have to consider the full expression to map accurately.
- We use multiple different contextual features for detection with accuracy of 90%.



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RESEARCH ARTICLE

- Detecting geospatial location descriptions in language text
- Kristin Stock^a, Christopher B. Jones ^b, Shaun Russell^a, Mansi and Niloofar Aflakia



AOPEN



Student: Chanthujan Chandrakumar

Investigators: Raj Prasanna, Caroline Holden, Max Stephens Amal Punchihewa, Marion Lara Tan

Stage 1: Selection of P-wave Detection Algorithm

•Our project commenced with a thorough selection process to identify an optimal Pwave detection algorithm. This choice is crucial as it accurately identifies P-wave arrivals within seismic waveforms. [1][2]

•We assessed multiple algorithms, considering their performance metrics and capacity to handle diverse seismic data. This comprehensive evaluation guarantees the reliability of our subsequent analysis. [3][4]





Stage 2: Building a relationship between P and S-wave's intensity

•Data Source Selection: Our study is centred on CUSP stations in Canterbury (2015-2022) with labelled P-wave picks, ensuring robust data quality.

•Waveform Collection: From the selected stations, we acquired 3085 waveforms, forming a substantial dataset for analysis.

•P-wave and S-wave Parameters: Extracted key parameters within three seconds of P-wave arrivals and during the S-wave using a tool (Figure 1) developed by the research team.

Relationship Building: Our ongoing work involves correlating the extracted parameters, and building an empirical relationship between P-wave and S-wave intensities (e.g., Figure 2).[5]



Modelling and Evaluation



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Key Challenges

Expanding Associate Investigator (AI) and Early Career Researcher capability through aligned opportunities and research programmes

- Jun 2023 IP4 Annual Meeting
- Establishment of QC Artificial Intelligence Group
- Running conferences to better connect researchers with stakeholders
 - Nov 2022 eGRID 7th IEEE Electronic Grid conference
 - Nov 2023 ISGT 12th IEEE PES innovative smart grid technologies conference
- Grant writing workshops for Endeavour Smart Ideas (4 applications x 2 Al's, \$1m grants)

Endeavour Programmes



[1] Transitioning Taranaki to a Volcanic Future \$16m, 2019-24 McDonald & Cronin

[2] Future Coast Aotearoa \$16m, 2021-26 Stephens, McDonald, et al.

[3] Adapting to Climate Change through Geothermal Enterprises \$8m, 2022-27 Cronin & McDonald

National Science Challenges



[1] Multi-hazard Risk Model \$4.5m, 2019-2024 Bebbington & McDonald



Questions

