

TEXAS A&M UNIVERSITY

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Hazard Reduction & Recovery Center

Overview

Introduction:

- Real estate transactions can be influenced by disasters and hazard maps. By analyzing changes in housing prices, planners could gain a deeper understanding of hazard information.
- Information on disaster impacts, proximity to hazards, and environmental cues are potentially associated with home-buying behaviors.
- However, disaster Hazard impacts on perception and prices fade over time, leading to price recovery.

Aim: Understand the individual and combined effects of explicit (maps) and experiential (events) landslide risk information on housing prices.

Methods:

- Area: Xizhi District, New Taipei City, Taiwan.
- Data: 39,836 housing transactions and their structural attribute data (2012-2024).
- Data Sources: Ministry of the Interior and Ministry of Agriculture, Taiwan.
- **Network Analysis:** To capture environmental attributes by route distance.
- **Model:** Hedonic Models and Difference-in-Differences Models.

Key Findings:

- Maps: Proximity to mapped risky streams negatively impacted prices (e.g., 37.64% drop within 100m).
- Actual Event: 2022 Landslide caused immediate and significant price drops (57.16%), but only on the older and weaker-structure house type (AB5F).
- **No Ripple Effect:** The landslide did not evoke further awareness and led to a price decrease near risky streams.
- **Temporal:** 2022 Landslide impact was short-lived (9 months), suggesting disaster myopia may exist.

Conclusion:

- Hazard maps can be effective guides and hence influence transactions. They can serve as mitigation information tools and , in combination with other measures (e.g., land-use, insurance), leverage development toward urban resilience.
- Disaster events can have severe but short-lived declines, indicating risk myopia. Windows of opportunity after disasters should be exploited to promote resilience, especially in the risky stream areas.

Research Hypotheses

H1:

Landslide hazard maps negatively affect the housing market; proximity to map-indicated risky areas is negatively associated with housing prices.



H2:

Actual landslide events cause additional declines in housing prices and amplify the negative effects in hazardous areas indicated on maps.



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H3:

Housing prices begin to recover over one year after landslide events.

Did Residents Move Away from Hazards? The Consequences of a Landslide and Landslide Risk Mapping for Housing Prices

Models

Model 1: Map Impact

 $ln_AdPrice = \beta_0 + \beta_1 S + \beta_2 E + \beta_3 HR$ Tests hazard map effects on housing prices

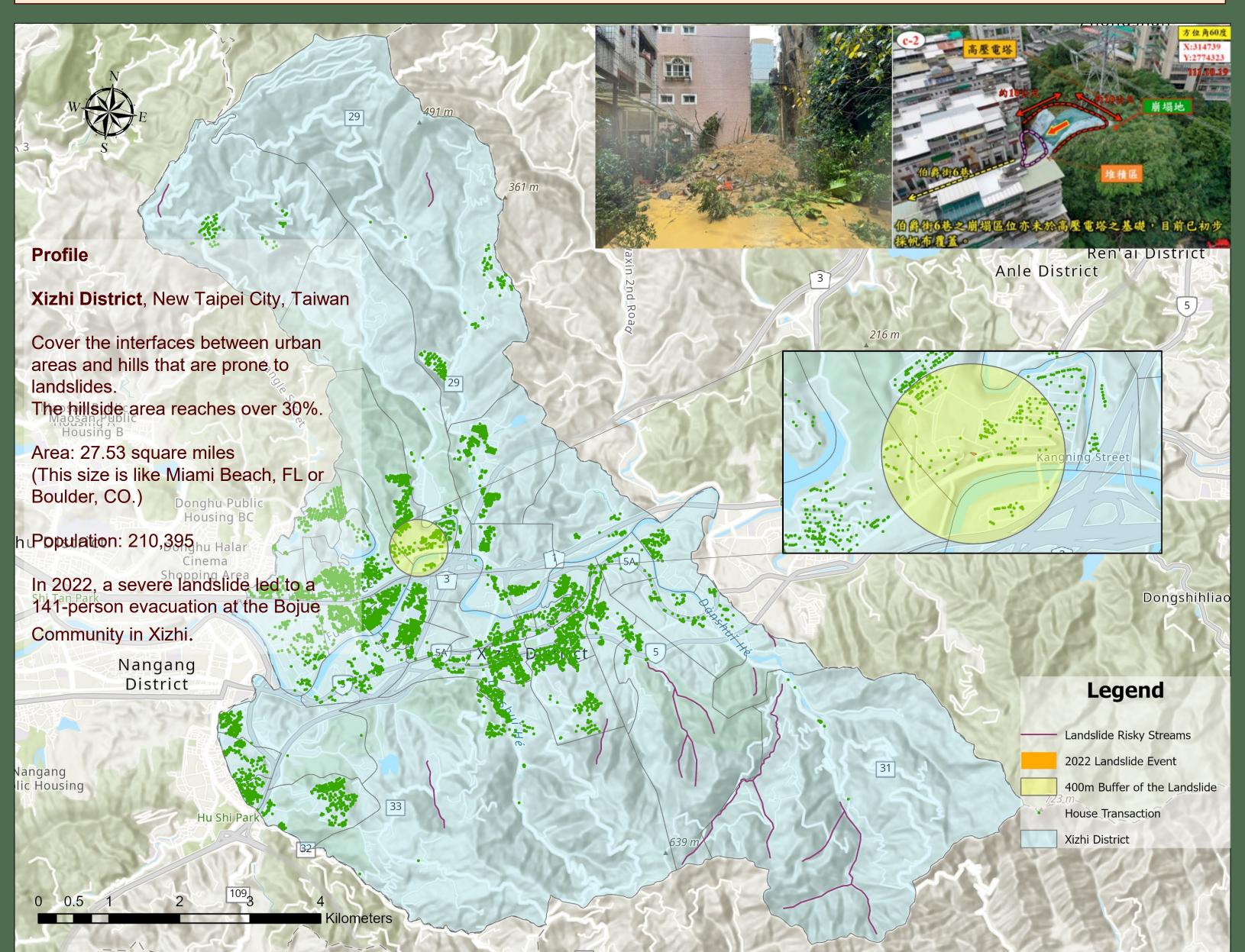
Model 2: Event Impact $ln_AdPrice = \beta_0 + \beta_1 S + \beta_2 E + \beta_3 Treat*Post + \alpha_i + \gamma_t$ Tests landslide event effects

Modified Model 2: Event Impact with Housing Types $ln_AdPrice = \beta_0 + \beta_1S + \beta_2E + \beta_3Treat*Post + \beta_4h_type + \beta_5Treat*Post*h_type + \alpha_i + \gamma_t$ *Tests event effects controlling for housing types*

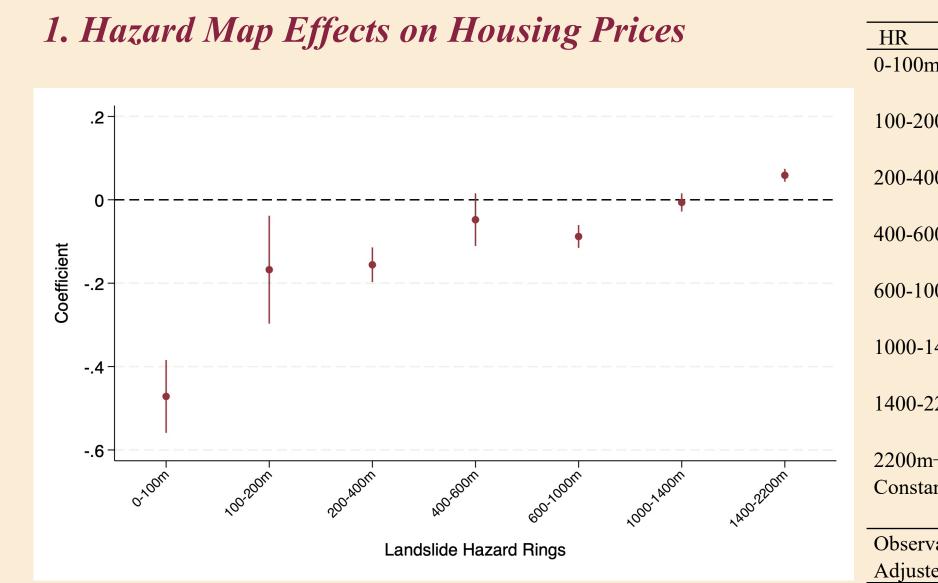
Model 3: Map-Event Interaction

In AdPrice = $\beta_0 + \beta_1 S + \beta_2 E + \beta_3 MTreat*Post + \alpha_i + \gamma_t$ *Tests interaction between map and event*

Variables: In AdPrice: Log Adjusted Price, S: Structural Attributes, E: Environmental Attributes, HR: Extent of Hazard Effect on the Map, h type: Housing Types, Treat: Interest Area of the Event, Mtreat: Interest Area of the Map, Post: After the Event, and TS: Time Series.



Results



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Model 4: Recovery Analysis ln AdPrice = $\beta_0 + \beta_1 S + \beta_2 E + \beta_3 TS$ *Tests temporal price recovery patterns*

	Log (Price)	
n	-0.472***	
	(0.045)	
)0m	-0.168*	
	(0.066)	
)0m	-0.156***	
	(0.021)	
)0m	-0.048	
	(0.032)	
)00m	-0.088***	
	(0.014)	
400m	-0.006	
	(0.011)	
2200m	0.059***	
	(0.008)	
1+	Base Group	
int	17.252***	
	(0.022)	
vations	39839	
ed R ²	0.526	

- Proximity to the mapped risky streams negatively impacted prices.
- The impact extends to 1,000 meters.
- Within 400-600 meters, there might be a trade-off between natural views and hazards.

Results

2. 2022 Land	slide Efj	fects on
Treat _{600m} *Post _{12mo} Treat _{600m} *Post _{9mo}	0.086 (0.044)	0.116* (0.054)
Treat _{600m} *Post _{6mo} Treat _{600m} *Post _{3mo}		
$Treat_{400m*}Post_{12mo}$ $Treat_{400m*}Post_{9mo}$		
$Treat_{400m*}Post_{6mo}$ $Treat_{400m*}Post_{3mo}$		
Observations	5314	3792
Adjusted R ²	0.472	0.534

	Single-family house	Apt. (11+ stories)	Apt. (6-10 stories)	Apt. (5- stories)	
Treat _{600m} *Post _{9mo}	+***	+	- (nonsignificant)	+***	 After controlling housing types, 5-story
Treat _{600m} *Post _{6mo}	+***	+***	- (nonsignificant)	-(nonsignificant)	high apartments (AB5F) received
Treat _{600m} *Post _{3mo}	+***	+	-(nonsignificant)	-(nonsignificant)	impacts after the landslide.
Treat _{400m*} Post _{12mo}	+***	+*	-(nonsignificant)	-(nonsignificant)	
Treat _{400m*} Post _{9mo}	+***	+**	+***	-*	 The impact sustained 9 months.
Treat _{400m*} Post _{6mo}	+***	+***	+***	_***	
Treat _{400m} *Post _{3mo}	+***	+***	na	_***	

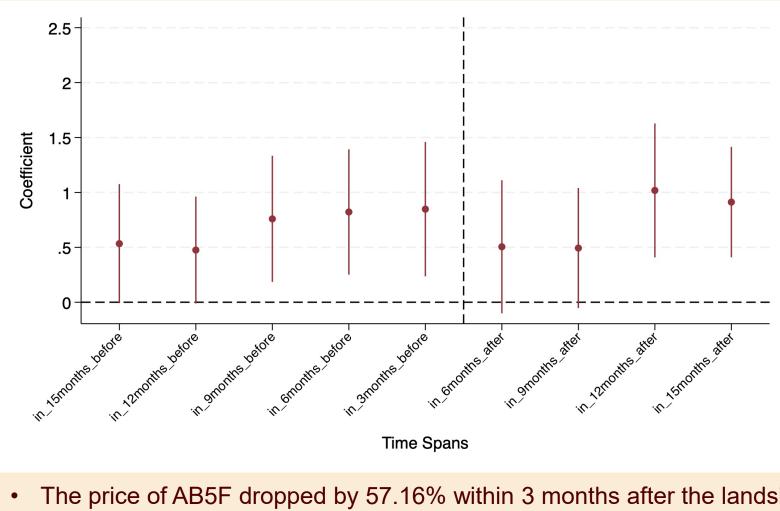
Log (Price)

4. Interaction between the Map and the Landslide

MTreat*Post _{12mo}	0.002	
	(0.028)	
MTreat*Post _{9mo}		-0.
		(0.0
MTreat*Post _{6mo}		
MTreat*Post _{2mo}		

2110		(0.022)		
MTreat*Post _{6mo}			-0.026	
			(0.031)	
MTreat*Post _{3mo}				-0.007
				(0.043)
Constant	18.114***	18.149***	17.088***	16.714***
	(0.553)	(0.551)	(0.598)	(0.595)
Observations	6009	4296	2652	1217
Adjusted R^2	0.439	0.451	0.522	0.577
5 Price Rec				0.577





• The price of AB5F dropped by 57.16% within 3 months after the landslide. AB5F took 9 months for housing prices to recover after the disaster

Implications in Hazard Mitigation Planning

- 4. Maintain accurate, updated hazard mapping systems

Housing Prices

• There is no significantly negative result without considering housing 0.126** types (0.045)• This model couldn't detect impacts 0.053 on the prices. (0.059) 0.084^{*} (0.032)0.098** (0.031)0.097 (0.028)(0.057)5314 3792 2324 1068 1068 0.565 0.551 0.472 0.534 0.565 0.551

3. 2022 Landslide Effects Controlling Housing Types on Housing Prices (β 3+ β 5)

Log (Price)

 There is no significant negative effect on houses near landslide-

prone streams after the disaster

TS	Log (Price)
15months_before	0.533
	(0.256)
12months_before	0.475
	(0.229)
9months_before	0.759*
	(0.271)
6months_before	0.822**
	(0.269)
3months_before	0.848**
	(0.289)
3months_after	Base Group
6months_after	0.505
	(0.286)
9months_after	0.493
	(0.258)
12months after	1.018**
_	(0.288)
15months after	0.912**
_	(0.237)
Constant	1.903
	(4.920)
Observations	35
Adjusted R2	0.609

1.Use housing prices to evaluate hazard map effectiveness

2.Implement market-based mitigation instruments based on hazard maps 3. Target interventions during post-disaster risk awareness peaks