Are seawalls correlated with lower property values in California?

Arlan Brucal, Qingran Li and John Lynham

June 2018

The Centre for Climate Change Economics and Policy (CCCEP) was established by the University of Leeds and the London School of Economics and Political Science in 2008 to advance public and private action on climate change through innovative, rigorous research. The Centre is funded by the UK Economic and Social Research Council. Its second phase started in 2013 and there are five integrated research themes:

1. Understanding green growth and climate-compatible development
2. Advancing climate finance and investment
3. Evaluating the performance of climate policies
4. Managing climate risks and uncertainties and strengthening climate services
5. Enabling rapid transitions in mitigation and adaptation

More information about the Centre for Climate Change Economics and Policy can be found at: www.cccep.ac.uk.

The Grantham Research Institute on Climate Change and the Environment was established by the London School of Economics and Political Science in 2008 to bring together international expertise on economics, finance, geography, the environment, international development and political economy to create a world-leading centre for policy-relevant research and training. The Institute is funded by the Grantham Foundation for the Protection of the Environment and the Global Green Growth Institute. It has six research themes:

1. Sustainable development
2. Finance, investment and insurance
3. Changing behaviours
4. Growth and innovation
5. Policy design and evaluation
6. Governance and legislation

More information about the Grantham Research Institute on Climate Change and the Environment can be found at: www.lse.ac.uk/grantham.

Are Seawalls Correlated with Lower Property Values in California?

Arlan Brucal[*] Qingran Li[†] John Lynham[‡]

This version: June 2018

Abstract

Rising sea levels necessitate careful consideration of different forms of coastal protection but cost-benefit analysis is limited when important non-market social costs have not been measured. Seawalls protect individual properties but can potentially impose negative externalities on neighboring properties via accelerated beach loss. We conduct a hedonic valuation of seawalls in two coastal California counties: San Diego and Santa Cruz. We do not find strong evidence to suggest that the presence of a seawall is positively correlated with the value of the home protected. However, we find that seawalls are strongly negatively correlated with the value of neighboring properties in Santa Cruz but not in San Diego, suggesting that the effect of seawalls depend on certain geographical attributes. Our results are robust even after accounting for the public-good nature of locational attributes and potential spatial dependence of housing prices. Simulation reveals that doubling the current level of seawalls in San Diego and Santa Cruz would reduce property tax revenues by \$7 million and \$54 million, respectively.

JEL-Classification: H23, Q26, Q51, Q54

Keywords: seawalls; hedonic valuation, coastal, erosion, property prices

1 Introduction

Rising sea levels necessitate careful consideration of the pros and cons of different forms of coastal protection. As sea levels rise, pressure to armor the coast, in the form of seawalls and other protective structures, will likely grow (Caldwell and Segall, 2007).[1]. However, putting up a seawall in front of one's property is a classic example of an action with a negative externality.[2] Seawalls protect individual properties but may impose negative externalities or costs on beachgoers and neighboring properties via accelerated beach loss. If such costs are greater than the private benefits, then seawalls might actually reduce social welfare.

How do seawalls cause accelerated beach loss? First, the physical area under the footprint of the actual seawall is lost. This is referred to as placement loss. Second, seawalls restrict the sediment and sand eroding down bluffs that would otherwise replenish and grow beaches. This is known as impoundment loss. Third, they magnify the power of retreating waves, carrying sand out to the sea. These outgoing reflected waves strike the sea floor on each outgoing surge, creating a steeper undersea slope.[3] These steeper slopes then preclude the ability of incoming waves to bring new sand to replenish an already eroded beach. The effect will be gradual loss of the beach in front of the seawall as the water deepens and the shoreface profile migrates landward. This process is known as passive erosion and has been well documented along many different shorelines (Griggs, 2005). Finally, seawalls often cause loss of nearby beaches. The new steep undersea slope prevents sand supply moving down the coast. As a result, neighboring beaches are starved of sand that would otherwise be supplied by shoreline currents (see Griggs and Tait, 1988; Hall et al., 1990 for detailed case studies and Griggs (2009) for an accessible overview).

It is well documented that people enjoy going to the beach (Bin et al., 2005) and that wide sandy beaches are highly valued (Pompe and Rinehart, 1995; Pompe, 1999; Landry et al., 2003;

---

[1]A number of studies show that sea level rise may increase the vulnerability of coastal areas to erosion and flood hazards, exposing more properties, facilities, and people to physical hazards and thus putting a pressure on their ability to cope with related disasters and to respond to them (Wu et al., 2002)

[2]We define a seawall as a vertical or near-vertical shore-parallel structure designed to prevent erosion and storm surge flooding.

[3]Figure A.1 in the appendix illustrates how coastal armoring accelerates beachloss.

Landry and Hindsley, 2011; Pendleton et al., 2012). Not surprisingly, better beach views, wider beaches, and closer proximity to the beach are all correlated with higher property valuation (Brown and Pollakowski, 1977; Edwards and Gable, 1991; Parsons and Wu, 1991; Parsons and Powell, 2001; Bin et al., 2008a) Some studies directly estimate the value of beach width in a hedonic framework and find a positive and significant relationship between beach width and property value (Pompe and Rinehart, 1995; Landry and Hindsley, 2011). Others estimate the reduction in property values due to higher erosion risk (Kriesel et al., 1993; Pompe and Rinehart, 1995; Bin et al., 2008b; Pompe, 2008). Because beach proximity and erosion risk are so highly correlated, separate identification within a hedonic framework is potentially challenging. In a major contribution to the literature, Bin et al. (2008a) construct a three-dimensional measure of ocean view that varies independent of erosion risk in order to separately identify these effects. They find that a greater ocean view significantly increases property value and flood risk significantly lowers property value. In a similar contribution, Gopalakrishnan et al. (2011) correct for potential endogeneity in hedonic models by instrumenting for beach width using spatially varying coastal features. They find that the beach width coefficient is nearly five times larger than the standard OLS estimate, suggesting that beach width is a much larger portion of property value than previously thought.

Despite this vast literature documenting the hedonic value of beach width and reduced erosion risk, it remains unclear whether seawalls themselves have a net negative or positive impact on social welfare. There are currently only two hedonic studies on the negative external cost of seawalls and/or other forms of coastal armoring. In the first, Kriesel and Friedman (2002, 2003) find that coastal armoring appears to lower property values a few rows inland. They find a negative and statistically significant correlation between property prices in the Southeast US and the degree of coastal armoring on nearby beaches. However, waterfront homes protected by a seawall tend to have higher values than waterfront properties that are unprotected. In the second study, Landry et al. (2003) conduct both a hedonic and a contingent valuation study of coastal armoring. They find a negative but statistically insignificant correlation between property prices on Tybee Island, Georgia, and a visible erosion structure at the nearest shore. Their contingent valuation study confirms that recreational beach visitors have a higher willingness to pay to visit beaches with minimal shoreline

armoring (approximately 25% higher). Thus, the literature to date strongly suggests that people who value beaches do not like beaches with seawalls but there is not an overwhelming amount of evidence that seawalls are correlated with lower property values.

Our goal in this paper is to test whether the presence of a seawall is correlated with higher or lower property values in coastal areas, holding all other determinants of value constant. Our paper is complementary to Kriesel and Friedman (2002, 2003) in that we use a very similar approach and data source, however, we use a different subset of the data from a different geographical location: California. In particular, we take data from San Diego and Santa Cruz counties, which offers variation in terms of the extent of spatial dependence in housing prices and difference in valuation of beach and coastal armoring arising from differences in certain geographical attributes, such as altitudes and distance to water.

In addition, we make a minor methodological improvement over the work of Kriesel and Friedman (2002, 2003) by estimating our model while testing and accounting for potential spatial dependence in the estimation procedure and the public good nature of locational attributes. Our work thus serves as both a replication test and an extension of Kriesel and Friedman's earlier findings. The rest of the paper is organized as follows. Section 2 presents the conceptual framework and the details of our empirical analysis. Section 3 presents the results and Section 4 discusses the implications of our findings.

2 Conceptual Framework

2.1 Hedonic valuation of seawalls

There are various factors that determine house values in coastal areas:

$$P_i = f(S_i, L_i, C_i, A_i) \tag{1}$$

4

where P is the price of a house i, S are structural attributes (such as bedrooms, square footage, lot size, etc), L are locational attributes (e.g. distance to water), C is the degree of coastal armoring on the nearest beach to the property (i.e. % of waterfront homes with seawalls) and A is an indicator variable that is equal to 1 if the property is waterfront and protected by a seawall. We will use a linear regression approach to determine how each attribute influences property values:

$$P_i = \alpha + N \sum_{j=1}^{N_s} \lambda_j S_{ji} + \sum_{l=1}^{N_L} \gamma_l L_{li} + \beta_1 C_i + \beta_2 A_i \varepsilon_i \tag{2}$$

where α is an intercept term, λ and γ are coefficients on the structural and locational attributes respectively, β_1 and β_2 are our coefficient of interest and ε is an error term with mean zero. If two houses, r and s are identical in S and L then:

$$P_r - P_s = \beta_i(C_r - C_s) + \beta_2(A_r - A_s) \tag{3}$$

is an empirically valid measure of the impact of coastal armoring on property values. This impact could be negative or positive, depending on the relative size of the private value to waterfront property owners who built seawalls and the value of seawall protection of other houses in their locality.

2.2 Data

The data employed to perform our analyses are from field measurement and mail survey information collected by the H. John Heinz III Center for Science, Economics and the Environment in 1999.[4] The Heinz Center published a report in 2000 entitled Evaluation of Erosion Hazards (Heinz Center, 2000). This report was mandated under the National Flood Insurance Reform Act of 1994, and involved a large team of collaborators whose goal was to assess coastal erosion and the resulting

---

[4]The dataset ia s result of a study commissioned by the Federal Emergency Management Agency (FEMA). No follow-up study assessments or household surveys were conducted after this study.

loss of property along the ocean and Great Lakes shorelines of the United States. There were many different components to this project but one element involved collecting data on the presence of seawalls and on property prices in coastal communities.

The data on seawalls and erosion risk was collected as follows. The Heinz Center sent field survey teams to measure and photograph 11,450 different structures in 18 coastal counties. The total number of surveyed structures in San Diego was 997 and 746 in Santa Cruz. The entire data set covers approximately 3% of the structures located within 1,000 feet of the shoreline along the Atlantic coast, Gulf of Mexico, Pacific coast, and the Great Lakes. Additional information on the surveyed structures was then obtained from county assessor offices. This was then supplemented with information from detailed questionnaires mailed to the owners of properties within 500ft of the shoreline within the 18 coastal counties. The Heinz Center research team combined all of this information to calculate the presence and extent of seawalls and beach nourishment projects within the 18 counties. The Heinz Center research team also collected data in the field on the reference elevation of each structure, beach width, and on the distance to an Erosion Reference Feature, typically the top edge of a bluff, dune, vegetation line or beach. This was done by a three-person team using GPS receivers. A complete list of variables and their description is provided in A.6.

The data on property prices and property characteristics was collected from the detailed questionnaire mentioned above.[5] This mail survey of coastal homeowners was conducted by the Heinz Center, in collaboration with the University of Georgia. In addition to questions regarding the presence of seawalls on local beaches, property owners were asked to report the purchase price of their home, the date purchased, and basic information about the property (number of bedrooms, square footage, etc.). This data was cross-referenced with data on property characteristics collected from county assessor and courthouse records.

The Heinz Center study is one of only a few detailed surveys of seawall locations in the US and is a unique resource for evaluating their impacts. Of the 18 counties studied in the original report,

---

[5]Ideally, property prices should be based on actual sales price. Without access to actual sales data, homeowner estimates of value can serve as reasonably good approximations (Taylor, 2003).

two are in California: San Diego and Santa Cruz counties. In this paper, we focus on the two California counties because previous research using the same dataset centered on the East coast of the U.S. and a California-specific analysis may be useful for informing policy discussions in the state regarding responses to sea level rise and other coastal hazards.

2.3 Hedonic pricing model

In terms of implementation, we apply a log-log hedonic pricing model:

$$ln(\text{Property Price}) = p_i = \beta_0 + \sum_{n=1}^{N} \beta_n ln X_n + \sum_{k=1}^{K} \beta_k D_k + \varepsilon \tag{4}$$

where all continuous independent variables (including the share of armored properties), denoted by X_n, are transformed into their natural logarithms, D_k represent dummy variables (including an indicator variable that turns to unity when the property is armored) and all other terms are as previously defined. We estimate equation 4 for the two counties (separately and combined) using simple ordinary least squares. Later on, we relax the assumptions of this method by (1) weighting locational attributes by lot size as proposed by Parsons (1990) to account for the public-good nature of locational attributes and (2) considering potential spatial dependence in housing prices (Cliff and Ord, 1970; Can, 1990).

It is expected that the seven structural characteristics variables, except the two temporal indicators, should have a positive influence on house prices. Age of house (*Years Since Built*) can capture the annual deterioration of the physical property and therefore should have a negative effect on house price. Due to price appreciation over this time period, recently sold houses should have higher prices than houses sold a few years ago, so we expect a negative sign on *Years Since Last Transaction*.

Other variables describe the environmental quality and risk of the structure. Coastal armoring projects (such as seawalls, rip-rap, groins, and breakwaters) are typically installed with the objective to protect properties from erosion damage. Therefore, the *Seawall* variable is expected to have

a positive coefficient in the regression model. For non-waterfront homes, erosion risk is usually not a consideration of the owners. Instead of benefiting from armoring projects, they are likely to worry more about the degradation of beach amenities from seawalls. For unprotected non-waterfront homes, reduced lateral access to shoreline due to coastal armoring can lower their property values (Griggs, 2009). This effect is captured by the variable *Percent Seawalls*, defined as the percent of waterfront structures within a beach community[6] that reported the presence of shoreline stabilization structures such as seawalls and rip-rap. This is our primary variable of interest and we expect the associated coefficient to be negative. This variable was calculated by dividing the sum of *Seawall* by the sum of *Waterfront* within each of the fifteen beach communities in our sample.[7]

Another shoreline protection option, beach nourishment, could also reduce erosion risk to coastal properties while maintaining some beach amenities. The two variables *Waterfront Nourishment* and *Beach Nourishment* are expected to have positive correlations with property prices. Respondents reporting beach nourishment only account for a small percentage of the California sample (9%). Thus, we observe no variation in this variable for the vast majority of houses in our sample. We therefore report results including and excluding those respondents who answered questions relating to beach nourishment.

Elevation refers to the height that the first floor of a home is elevated above the base flood elevation level. The higher elevation of the house, the better the physical flood protection. *Geotime* is a measure of erosion risk defined as the number of years until the distance between the property and the erosion reference feature (ERF) is reduced to zero. To compute this variable, the erosion rate at each property location is calculated first. Since in some areas erosion rates can be determined from historical records, while in others they are affected by protection devices such as seawalls, we applied the same method used in the Heinz Centers erosion hazard report. A detailed description of this formulation is presented in Appendix 4. A squared logarithm of *Geotime* is added to the

---

[6]Our dataset contains nine San Diego beach communities and six Santa Cruz beach communities, with the largest community having 50 property records. A beach community is defined as all of the buildings located within 1,000 feet of the shoreline.

[7]Ideally, the measure of the extent of seawalled coastline should be the total length of seawalls divided by the total length of the coastline in a particular beach community. However, our dataset does not contain the length of seawalls making this metric infeasible to calculate.

model to test for a non-linear effect. *Built Post-FIRM* is an indicator variable for whether or not the home was built post the publication of a Flood Insurance Rate Map (FIRM) for the surrounding area. Post-FIRM homes typically have lower insurance premiums if they are built above base flood elevation levels, therefore we expect the coefficient on this variable to be positive.

Table 1 presents summary statistics for all of our main variables. On the average, properties in Santa Cruz are generally more expensive on the average compared to those in San Diego. This is probably due to bigger house and lot sizes. It is also quite noticeable that the houses in San Diego are situated in higher places and closer to the water, which implies that these houses are potentially in high cliffs above water (Griggs and Savoy, 1985). The difference in this locational attribute should provide more insights on the effect of seawalls on inland property. In particular, seawalls in Santa Cruz should affect inland properties much more than they could in San Diego. The extent of seawalled properties are about the same for the two counties, which makes these two cases comparable.

3   Results

3.1   OLS results

We start our analysis with an ordinary least squares (OLS) method using all the variables listed in Table A.6. We did separate analysis for San Diego and Santa Cruz to account for potential difference in hedonic price functions between these two market segments. We also run combined observations from the two counties to formally test this difference by adding a *San Diego* indicator variable that turns on when the property is situated in San Diego County. Results are summarized in Table 2.

For San Diego, we reject the null hypothesis of no correlation for only 4 variables. In particular, we find that price increases by about 0.50% as the size of a house increases by one percent.[8] This

---

[8]It is extremely important to note that all of our regression coefficients should be interpreted as correlations and

Table 1: Summary Statistics.

	San Diego		Santa Cruz	
	Mean	SD	Mean	SD
Price ('000, in 2014 US$)	588.56	412.20	699.08	600.85
Seawall (0-1)	0.18	0.38	0.19	0.39
Percent seawalled (%)	39.59	17.79	42.49	21.00
House size (sq. ft.)	1784.05	811.17	2571.56	2377.86
Lot size (sq. ft.)	5843.64	5019.31	8032.56	4873.24
Years since last transaction	14.42	10.52	18.62	12.84
Years since built	40.09	19.86	44.78	20.56
Fireplace (0-1)	0.81	0.39	0.72	0.45
Waterfront (0-1)	0.47	0.50	0.44	0.50
Waterfront nourishment (1-0)	0.06	0.24	0.02	0.12
Beach nourishment (1-0)	0.13	0.34	0.05	0.21
Bedroom	2.93	1.09	3.19	1.47
Geotime	2289.42	6609.13	4895.74	8636.80
Elevation (ft.)	7801.26	4220.72	1327.22	3360.50
Distance to water (ft.)	186.75	162.45	249.77	161.57
Beach width	81.21	286.96	38.39	179.09
Built Post-FIRM	0.05	0.21	0.07	0.25
Observations	132		151	

Source of basic data: Heinz Center (2000).

translates to a marginal implicit price of house size by about \$164/sq.ft.[9] Lot size has smaller marginal implicit price, with houses being valued more by about \$37 as lot size increases by sq.ft. Waterfront is surprisingly not significant. This is possibly due to fact that most houses in San Diego are situated in high cliffs above water and beach is not easily accessible. Surprisingly, homeowners value having a fireplace in their homes, considering that most areas of San Diego rarely require a fireplace for heating. This is possibly because we did not explicitly know whether it is outdoor or indoor fireplaces. Outdoor fireplaces may be associated with homeowners' valuation of outdoor

not as causal effects. See Gopalakrishnan et al. (2011) for discussion of the endogeneity problem in hedonic beach value models and for a potential solution.

[9]We calculate marginal implicit price (\$ per unit) for the logged variables as:

$$\text{Marginal Implicit Price} = \frac{\beta_k}{\bar{x}_k} * \bar{p}$$

where β_k is the estimated coefficient of the regressor x_k, and $\bar{x}_k$ and $\bar{p}$ are the unconditional means of the regressor x_k and the dependent variable p, respectively.

living, which makes this structural attribute a worthy investment.[10]

---

[10]The National Association of Realtors 2013 Home Features Survey reveals that 57% chose homes with at least one fireplace.

Table 2: Results from simple OLS.

	All Variables			Reduced Model			Reduced Model		
Regression	San Diego	Santa Cruz	Combined	San Diego	Santa Cruz	Combined	San Diego	Santa Cruz	Combined
Seawall (0-1)	0.216 (0.182)	0.100 (0.206)	0.099 (0.130)	0.223 (0.159)	0.068 (0.157)	0.114 (0.110)	0.216 (0.157)	-0.007 (0.146)	0.079 (0.107)
Percent seawalled (%)	-0.038 (0.045)	-0.111** (0.050)	-0.080*** (0.030)	-0.061 (0.038)	-0.100*** (0.028)	-0.091*** (0.024)	-0.065* (0.038)	-0.094*** (0.028)	-0.088*** (0.024)
House size (sq. ft.)	0.498*** (0.125)	0.523*** (0.163)	0.516*** (0.113)	0.455*** (0.105)	0.480*** (0.131)	0.496*** (0.089)	0.460*** (0.104)	0.504*** (0.129)	0.494*** (0.088)
Lot size (sq. ft.)	0.370*** (0.097)	0.057 (0.073)	0.150** (0.061)	0.373*** (0.089)	0.079 (0.065)	0.133** (0.053)	0.375*** (0.089)	0.076 (0.063)	0.137*** (0.053)
Years since last transaction	-0.392*** (0.058)	-0.375*** (0.057)	-0.387*** (0.040)	-0.379*** (0.055)	-0.355*** (0.048)	-0.374*** (0.036)	-0.380*** (0.055)	-0.359*** (0.047)	-0.372*** (0.036)
Years since built	-0.134 (0.125)	-0.218 (0.174)	-0.174* (0.095)	-0.050 (0.099)	-0.106 (0.085)	-0.064 (0.065)	-0.052 (0.097)	-0.102 (0.087)	-0.070 (0.065)
Fireplace (0-1)	0.267*** (0.096)	0.181 (0.115)	0.232*** (0.075)	0.255** (0.105)	0.187* (0.104)	0.213*** (0.074)	0.241** (0.103)	0.191* (0.104)	0.211*** (0.074)
Waterfront (0-1)	0.165 (0.119)	0.333** (0.159)	0.290*** (0.106)	0.196 (0.134)	0.463*** (0.117)	0.375*** (0.088)	0.207 (0.133)	0.428*** (0.112)	0.365*** (0.086)
Waterfront nourishment (1-0)	0.147 (0.620)	-0.216 (0.234)	-0.147 (0.232)	0.118 (0.582)	-0.447** (0.204)	-0.270 (0.205)			
Beach nourishment (1-0)	-0.001 (0.456)	0.124 (0.168)	0.056 (0.180)	-0.205 (0.408)	0.209** (0.104)	0.087 (0.137)			
Bedrooms	-0.066 (0.116)	0.032 (0.137)	-0.036 (0.088)						
Geotime	-0.034 (0.322)	-0.001 (0.076)	0.008 (0.066)						
Geotime squared	-0.005 (0.021)	-0.000 (0.007)	-0.005 (0.005)						
Elevation (ft.)	-0.004 (0.027)	0.020 (0.023)	0.013 (0.016)						
Distance to water (ft.)	0.027 (0.109)	-0.023 (0.049)	-0.009 (0.043)						
Beach width	0.019 (0.031)	0.013 (0.029)	0.015 (0.021)						
Built Post-FIRM	-0.580 (0.356)	-0.476 (0.401)	-0.540** (0.239)						
San Diego (1-0)			0.132 (0.104)			-0.015 (0.076)			-0.013 (0.075)
Constant	7.889*** (1.402)	10.385*** (1.132)	9.587*** (0.767)	7.445*** (0.863)	10.067*** (0.961)	9.380*** (0.652)	7.414*** (0.866)	9.913*** (0.974)	9.374*** (0.653)
Observations	130	117	247	130	151	281	130	151	281
R-sq.	0.623	0.550	0.557	0.574	0.497	0.516	0.572	0.486	0.513
R-sq. (adj.)	0.566	0.473	0.522	0.538	0.461	0.496	0.544	0.457	0.497

Note: Robust standard errors are in parentheses. *, **, *** denote statistical significance at 10, 5 and 1 percent level.

In Santa Cruz, house size is positively correlated with housing prices. However, since houses in Santa Cruz are larger on the average, the implicit marginal price of house size is smaller at \$135/sq.ft. The same observation holds true for lot size, with Santa Cruz houses being valued by \$32 more per additional sq.ft. *Waterfront*, in contrast to San Diego, is significantly associated with higher housing prices in Santa Cruz. In particular, waterfront houses in Santa Cruz are valued more by about 40%, posting a marginal implicit price of more than \$232,000.[11] The existence of fireplace, while positive, is not significantly correlated with housing prices.

Overall, we find no strong evidence to suggest that there is systematic differences in housing prices between the two counties or market segmentation Taylor (2003). Our OLS regression finds that the San Diego dummy, while positive, is statistically insignificant. We also performed Wald test for San Diego dummy and found the F-statistic insignificant, with the value of 1.60 and p-value equivalent to 0.207.

The variables indicating flooding and erosion risks (e.g. *Geotime*, *Elevation*) do not appear to be significantly correlated with property prices. This is true when we analyze housing prices in each county separately and after combining observations from the two counties. This is different from the findings in (Kriesel and Friedman, 2003). This could potentially be explained by the different nature of the shorelines between the two regions. Unlike the Atlantic and Gulf coast, much of the Pacific coastline consists of narrow beaches backed by steep cliffs. Cliff erosion is site specific and episodic, which is often hidden by the low long-term average erosion rate. Only 45 observations in our entire sample have a *Geotime* less than 100 years. It is likely that erosion risk is not reflected in property prices till decades later.

Although the low long-term erosion rate might reduce homeowners' perceptions of the erosion threat to their properties, shoreline management policies can still affect house buyers aware-ness of erosion risk levels. Regarding the seawall indicators, we obtain remarkably similar re-

---

[11]For dummy variables, the calculated marginal implicit price is expressed as:

$$\text{Marginal Implicit Price} = (e^{\beta_k} - 1) * \bar{p} \tag{5}$$

where β_k and p are as previously defined.

sults to Kriesel and Friedman (2003); Kriesel et al. (2005). The coefficient on *Seawall* is positive but not statistically significant in either county. The imprecision of this estimate relative to Kriesel and Friedman (2003) could be due to our smaller sample size. We evaluate the effect of seawalls on housing prices using combined samples to increase the sample size and observe same result.

Meanwhile, we notice a significant negative effect of *Percent Seawalls* on property prices in Santa Cruz. This suggests that property owners in the county are worse off as a result of extensive coastal armoring. Furthermore, our coefficient of -0.11 is more than three times larger than the -0.03 found by Kriesel and Friedman (2002). This suggests that, as the percentage of waterfront coastal armoring in a community increases by 100%, the associated value of a house located on the Santa Cruz coast drops by $77,597. In contrast, we do not find any strong evidence to suggest that the extent of armored coastline in San Diego has an effect on property values. If there is an effect, house values in San Diego declines by about $22,365 if the share of waterfront coastal armoring in a community increases by 100%. Interestingly, our estimate falls within the ball park of Kriesel and Friedman (2003) who found that for the same percentage increase in coastal armoring, house prices on the Atlantic and Gulf coasts drop by $41,091 in 2014 dollars.

To analyze the net effect of building a seawall from a point of view of a marginal house, we need to account for the joint effect of building a seawall in front of the property and the added armored coastline on housing prices. Figure 1 summarizes the results using estimates from Table 2. The solid green line reflects the change in the property value of the marginal household from building a seawall, given an assumed change in the share of armored waterfront properties in the county. In San Diego, even if a group of coastal households decides to build seawalls on their properties, bringing the share of armored properties to 80%, we will not find any statistically significant effect on their housing prices. In contrast, we find a significant drop in the value of a marginal house in Santa Cruz if building a seawall would result in an increase in armored waterfront properties by just about 9% (see vertical red line). The drop in the property resale values is about 1% and can go up to 10% if building seawalls would result in increasing the current share of protected waterfront

14

properties from 42% to 84%.

Figure 1: Moran scatterplots for Housing Prices in San Diego and Santa Cruz counties
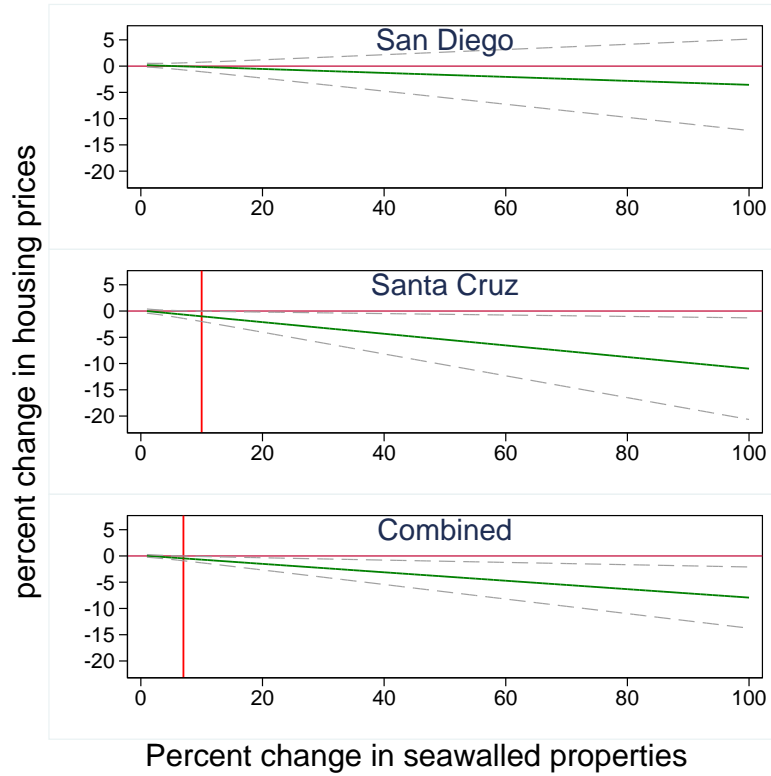
Source: Authors' estimates.

We also ran two OLS models with a smaller subset of variables in order to omit variables that seemed unimportant and increase our sample size slightly for Santa Cruz. The results of these reduced models are presented in columns 5 through 10 in Table 2. The coefficients that are statistically significant in the reduced models are mostly consistent with those in the complete model. The coefficient on *Seawall* is still not statistically significant. The coefficient on *Percent Seawalls* is about the same level as shown in Table 2 for Santa Cruz but larger for San Diego and marginally significant when remove variables relating to nearby and front beach nourishments.

We also consider the world suggested by Parsons (1990) in which locational amenities are pure public goods and as a result the hedonic function must have such attributes weighted by lot size.

This setting is quite applicable to our analysis because the number of individuals that may enjoy the benefits of a beach of certain quality is limited by the area of land over which that amenity is available. For example, there is a fixed amount of land within a mile from the Natural Bridges State Beach in Santa Cruz. Thus, the more land is taken by one household in a nearby location, the less available to others the benefits associated with living nearby the beach. Parsons (1990) argues that the size of this lost opportunity for others is proportional to lot size occupied by the household.

To account for the pure public good nature of location attributes, we modified equation 2 by using the log-transformed price per unit of land as the dependent variable. This procedure is equivalent to a weighting rule suggested by Parsons (1990), which allows locational attribute prices to vary proportionally with the amount of land occupied. Results using this method are shown in rows 2-4 of Table A.7. The coefficients of our variables of interest, namely *Seawall* and *Percent Seawall*, are comparable to the baseline result. We also formulate hedonic price regressions without the structural attributes similar to (Diamond, 1980) (see rows 5-7 of Table A.7). Our results remain robust and similar to the baseline results.

3.2   Accounting for spatial dependence

Testing for the presence of spatial autocorrelation in studying housing prices is important principally because the presence of spatial dependence violates one of the major properties of regression analysis – the independence across observations (Can, 1990). Ignoring spatial autocorrelation can lead to biased parameter estimates and misleading significance levels. In our context, if the prices of nearby houses have an absolute effect on each other, potentially due to the fact that some sellers or buyers assess a given property not only based on location but also on prices of nearby houses, then we may need to add spatial autoregressive terms in our specification. Equation 2 becomes a model with N cross-section units for each $i \in N$ as specified below:

$$P_i = \lambda \mathbf{WP} + \mathbf{X}_i \beta + \varepsilon_i \tag{6}$$

with

$$\varepsilon = \rho \mathbf{W} \varepsilon + u$$

where $\mathbf{W}$ is an nxn inverse distance weight matrix, $\mathbf{WP}$ is a vector of the spatially lagged house prices and λ and ρ are the unknown spatial lagged paramaters. $\mathbf{X}$ is a vector of explanatory variables, including *Seawall* and *Percent Seawall*, and u is a standard i.i.d. error term.

We account for the presence of spatial autocorrelation as specified in alternative forms. In particular, we test for the following hypotheses: (1) $\lambda = 0, \rho \neq 0$ which leads to a spatial lag model (SLM); $\lambda \neq 0, \rho = 0$ leading to spatial errors model (SEM); and (3) $\lambda \neq 0, \rho \neq 0$ leading to a spatial ARAR(1,1) model (Arraiz et al., 2010).

3.2.1 Tests for Spatial Dependence

A convenient and perhaps the most transparent way of testing for spatial dependence is in terms of graphical representation. Figure 2 shows the Moran scatterplot of house prices for San Diego and Santa Cruz counties. Moran scatterplots show the relationship between the individual house's price (in standardized form) in the horizontal axis and the average prices of neighboring houses weighted by the spatial lag vector $\mathbf{W}$ in the vertical axis. We assume that the degree of spatial dependence of a particular house is inversely related to the geographic distance between the two properties. That is,

$$\mathbf{W} = \begin{pmatrix} 0 & w_{1,2} & w_{1,3} & \dots & w_{1,n} \\ w_{2,1} & 0 & w_{2,3} & \dots & w_{2,n} \\ w_{3,1} & w_{3,2} & 0 & \dots & w_{3,n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ w_{n,1} & w_{n,2} & w_{n,3} & \dots & 0 \end{pmatrix} \tag{7}$$

$$w_{i,j} = \frac{1}{d_{ij}}$$

17

We find evidence of positive spatial dependence for houses located in Santa Cruz. The Moran statistic is 0.0867 and is statistically significant at 5-percent level. In contrast, there is no strong evidence to suggest that housing prices in San Diego are spatially dependent.

Figure 2: Moran scatterplots for Housing Prices in San Diego and Santa Cruz counties
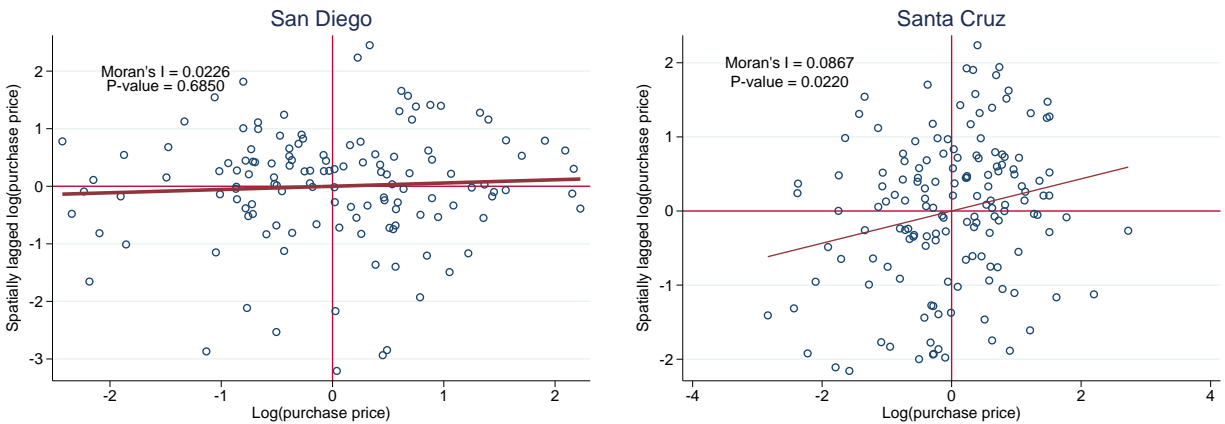
Source of basic data: Heinz Center (2000).

We formally test for the presence of global spatial autocorrelation by calculating Moran's I (Moran, 1950) and Geary c (Cliff and Ord, 1970). We then determine the significance of these statistics following the procedure suggested by Pisati (2001). Table 3 summarizes the results. The tests confirm our graphical approach suggesting no strong spatial dependence of house prices in San Diego. In contrast, there is strong evidence of spatial autocorrelation for Santa Cruz. The spatial autocorrelation persists even if we combined the two datasets.

3.2.2 Hedonic prices with spatial dependence

With an indication of potential spatial dependence, we re-estimate equation 2 using data for Santa Cruz county. This time, we relax our assumption that λ and ρ are both zero. We estimate models with spatial dependence in a maximum likelihood framework using the procedure suggested by Drukker et al. (2015). We employ different specifications to potentially account for different forms of spatial autocorrelation, namely spatial lag model, spatial error model and the spatial

Table 3: Tests for global spatial autocorrelation.

	San Diego	Santa Cruz	Combined
Distance of nearest neighbor (in kms)			
Mean	1.279	0.042	
Min	0.931	0.006	
Max	3.336	0.239	
Moran's I	0.023	0.087**	0.054*
	(0.049)	(0.037)	(0.022)
Geary c	0.943	0.900**	0.924**
	(0.051)	(0.040)	(0.024)

Notes: The estimated standard deviation of each statistic are in parentheses. *,**,*** represents significance at 10, 5 and 1 percent, respectively.

ARAR(1,1,) model. Then, for each model, we run three different specifications similar to the simple OLS regression in which we start with all the variables and drop some of them to slightly increase our sample. We do this to test the stability of our estimates. Results are presented in Table 4.

We start our analysis by looking at the estimated spatial autocorrelation parameters. We find no strong evidence to suggest that there exists spatial autocorrelation. In particular, we fail to reject the null hypotheses that (1) $\lambda = 0, \rho \neq 0$; (2) $\lambda \neq 0, \rho = 0$; and (3) $\lambda \neq 0, \rho \neq 0$ in equation 6.[12]

The estimated effect of changes in the variables of interest, even after accounting for potential spatial dependence, remains robust and consistent with our baseline results. The coefficient remain insignificant for the indicator variable *Seawall*. Nonetheless, the extent of coastline armoring (*Percent Seawalls*), remain negative albeit slightly smaller and statistically significant. This confirms our earlier result that as the percentage of waterfront coastal armoring in a beach community increases, the associated value of houses located in San Diego and Santa Cruz drops significantly. Moreover, our estimates remain stable across different forms of assumed spatial dependence and under different sets of controls, thus increasing our confidence on the robustness of the results.

---

[12]We also perform diagnostic tests for spatial dependence in an OLS framework after estimating equation 2, following the procedure as suggested by Pisati (2001). Results are summarized in Table A.8. We find no strong evidence to suggest that there is spatial dependence amongst observed housing prices in each county.

Table 4: Results for models with spatial dependence.

	Spatial Lag Model			Spatial Error Model			Spatial ARAR(1,1) Model		
	(1)	(2)	(3)	(1)	(2)	(3)	(1)	(2)	(3)
Seawall (0-1)	0.094	0.068	-0.008	0.103	0.060	-0.011	0.084	0.046	-0.026
	(0.161)	(0.138)	(0.128)	(0.163)	(0.136)	(0.126)	(0.157)	(0.132)	(0.123)
Percent seawalled (%)	-0.098*	-0.100**	-0.093**	-0.112**	-0.100***	-0.094**	-0.081*	-0.090**	-0.084**
	(0.053)	(0.041)	(0.041)	(0.054)	(0.037)	(0.038)	(0.047)	(0.035)	(0.035)
House size (sq. ft.)	0.496***	0.480***	0.502***	0.519***	0.484***	0.508***	0.472***	0.468***	0.489***
	(0.155)	(0.117)	(0.118)	(0.155)	(0.114)	(0.115)	(0.149)	(0.112)	(0.113)
Lot size (sq. ft.)	0.056	0.079	0.076	0.056	0.073	0.071	0.064	0.064	0.062
	(0.075)	(0.060)	(0.059)	(0.077)	(0.058)	(0.058)	(0.070)	(0.055)	(0.055)
Years since last transaction	-0.379***	-0.355***	-0.360***	-0.374***	-0.358***	-0.363***	-0.382***	-0.363***	-0.367***
	(0.057)	(0.053)	(0.053)	(0.058)	(0.053)	(0.053)	(0.057)	(0.053)	(0.052)
Years since built	-0.196	-0.106	-0.100	-0.210*	-0.113	-0.107	-0.212*	-0.102	-0.097
	(0.124)	(0.080)	(0.081)	(0.127)	(0.079)	(0.080)	(0.125)	(0.078)	(0.079)
Fireplace (0-1)	0.176	0.187*	0.191*	0.183	0.177	0.184*	0.142	0.165	0.171
	(0.116)	(0.108)	(0.108)	(0.117)	(0.108)	(0.108)	(0.119)	(0.108)	(0.108)
Waterfront (0-1)	0.333**	0.463***	0.426***	0.326**	0.480***	0.441***	0.371***	0.470***	0.434***
	(0.131)	(0.101)	(0.099)	(0.136)	(0.100)	(0.098)	(0.133)	(0.096)	(0.094)
Waterfront nourishment (1-0)	-0.208	-0.446*		-0.216	-0.458*		-0.200	-0.453*	
	(0.263)	(0.251)		(0.264)	(0.252)		(0.264)	(0.251)	
Beach nourishment (1-0)	0.132	0.209		0.127	0.220		0.128	0.196	
	(0.207)	(0.166)		(0.208)	(0.163)		(0.203)	(0.163)	
Geotime	0.033			0.028			0.054		
	(0.134)			(0.136)			(0.132)		
Geotime squared	0.021			0.010			-0.007		
	(0.080)			(0.093)			(0.076)		
Elevation (ft.)	-0.002			-0.001			-0.000		
	(0.006)			(0.007)			(0.006)		
Distance to water (ft.)	0.027			0.020			0.036*		
	(0.021)			(0.020)			(0.020)		
Beach width	-0.033			-0.030			-0.009		
	(0.049)			(0.059)			(0.049)		
Built Post-FIRM	0.010			0.013			0.005		
	(0.022)			(0.022)			(0.023)		
San Diego (1-0)	-0.426			-0.449			-0.512*		
	(0.290)			(0.312)			(0.302)		
Constant	8.372***	10.034***	9.640***	10.398***	10.114***	9.959***	6.603***	8.362***	8.102***
	(2.221)	(1.904)	(1.895)	(1.043)	(0.803)	(0.809)	(2.470)	(2.122)	(2.077)
Lambda	0.156	0.003	0.022				0.294	0.146	0.155
	(0.152)	(0.136)	(0.135)				(0.181)	(0.166)	(0.162)
Rho				0.056	-0.134	-0.107	-0.341	-0.312	-0.294
				(0.250)	(0.204)	(0.201)	(0.339)	(0.287)	(0.280)
Observations	117	151	151	117	151	151	117	151	151

Note: Robust standard errors are in parentheses. *, **, *** denote statistical significance at 10, 5 and 1 percent level.

4 Discussion

We find that the sales price of a house located within 1,000 feet of the coast in San Diego and Santa Cruz counties is correlated with its structural characteristics and the shoreline management policies taken in the community. Although seawalls protect coastal structures from erosion damages, the negative coefficient on *Percent Seawalls* calls for attention to the welfare impacts of coastal armoring on waterfront and non-waterfront households.

Using our baseline estimates, we predict the change in total housing values for the two counties when the share of those that are protected by seawalls increases by a certain percentage (see Table 5). Results suggest that doubling the current level of properties protected by seawalls is associated with a \$20,954.16 drop in the resale value of an average coastal home in San Diego that is protected by seawalls. For those that are not protected by seawalls, the magnitude is larger at \$22,365. Overall, the said increase in the share of protected properties translates to more than \$718 million reduction in the value of affected houses in the county. In terms of property tax revenues, a very crude back-of-the-envelope calculation suggests that this increase would cost the government \$7 million in terms of revenue lost and the discounted present value of this annual loss is \$143 million.[13]

For Santa Cruz County, the potential effect of increasing the current share of seawalled properties by 100% is significantly greater, reaching to about a \$77,000 decline in home resale values for protected properties. For non-seawalled properties, the effect is about similar at \$78,000. This implies that there is no positive net benefit of putting up an additional seawall from the current level. The increase in the seawalled-properties is associated with the loss in the total value of coastal homes in the county by more than \$5.4 billion. This translates to a government revenue lost of about \$54 million, which is equivalent to a discounted present value of about \$1.08 billion.

Although our estimates are large, the associated welfare impacts are consistent with the existing

---

[13]Property tax in San Diego County is a little more than 1%. Using the annual discount rate of 5%, an infinite stream of annual losses of \$7.188 million would be \$7.188m/0.05 = \$143.765m.

Table 5: Potential effect of increasing the share of waterfront households protected by seawalls.

San Diego						
% Change in Seawalls	Change in Price		No. of affected houses		Total effect	Revenue Lost
(from current mean)	(in $)	(in $)	seawalled	non-seawalled	(in '000$)	(in '000$)
1.00	10.47	-223.65	14,397	19,542	-4,219.92	-42.20
5.00	7.90	-1,118.26	14,967	18,972	-21,097.37	-210.97
10.00	-95.32	-2,236.53	15,680	18,259	-42,331.75	-423.32
20.00	-635.10	-4,473.06	17,105	16,834	-86,161.88	-861.62
30.00	-1,619.35	-6,709.58	18,531	15,408	-133,391.07	-1,333.91
40.00	-3,048.07	-8,946.11	19,956	13,983	-185,919.98	-1,859.20
50.00	-4,921.25	-11,182.64	21,382	12,557	-245,649.31	-2,456.49
60.00	-7,238.90	-13,419.17	22,807	11,132	-314,479.74	-3,144.80
70.00	-10,001.02	-15,655.70	24,232	9,707	-394,311.94	-3,943.12
80.00	-13,207.60	-17,892.22	25,658	8,281	-487,046.53	-4,870.47
90.00	-16,858.64	-20,128.75	27,083	6,856	-594,584.26	-5,945.84
100.00	-20,954.16	-22,365.28	28,509	5,430	-718,825.86	-7,188.26

Santa Cruz						
% Change in Seawalls	Change in Price		No. of affected houses		Total effect	Revenue Lost
(from current mean)	(in $)	(in $)	seawalled	non-seawalled	(in '000$)	(in '000$)
1.00	-0.74	-775.98	2,840.12	4,189.88	-32,533.71	-325.34
5.00	-158.54	-3,879.89	2,952.60	4,077.40	-162,879.83	-1,628.80
10.00	-704.13	-7,759.79	3,093.20	3,936.80	-327,267.40	-3,272.67
20.00	-2,956.45	-15,519.58	3,374.40	3,655.60	-667,096.04	-6,670.96
30.00	-6,756.96	-23,279.37	3,655.60	3,374.40	-1,032,546.40	-10,325.46
40.00	-12,105.67	-31,039.15	3,936.80	3,093.20	-1,436,679.04	-14,366.79
50.00	-19,002.57	-38,798.94	4,218.00	2,812.00	-1,892,554.72	-18,925.55
60.00	-27,447.66	-46,558.73	4,499.20	2,530.80	-2,413,233.60	-24,132.34
70.00	-37,440.95	-54,318.52	4,780.40	2,249.60	-3,011,776.64	-30,117.77
80.00	-48,982.43	-62,078.30	5,061.60	1,968.40	-3,701,244.16	-37,012.44
90.00	-62,072.11	-69,838.09	5,342.80	1,687.20	-4,494,696.64	-44,946.97
100.00	-76,709.97	-77,597.88	5,624.00	1,406.00	-5,405,194.88	-54,051.95

Notes:

I. San Diego County has 70 miles of coastline. Based on a very conservative estimate of housing density (4 homes per acre), this means that there are approximately 33,939 homes within 1,000 feet of the shoreline. 42% of these homes are currently protected by seawalls Property tax in San Diego County is a little more than 1%.

II. Santa Cruz County has 29 miles of coastline. Based on an even more conservative estimate of housing density (2 homes per acre), this means that there are approximately 7,030 homes within 1,000 feet of the shoreline. 40% of these homes are currently protected by seawalls. Property tax in Santa Cruz County is a little more than 1%.

III. The change in price for seawalled properties is calculated using the estimated coefficients of the *Seawall* and *Percent Seawall* in our baseline model (i.e., OLS including all variables). The unconditional mean property price for each county is used as the base price ($588,000 for San Diego and $699,000 for Santa Cruz.). For non-seawalled properties, we just used teh coefficient for *Percent Seawall*.

literature. Our findings are in line with Kriesel and Friedman (2003); Kriesel et al. (2005) who find negative, albeit smaller, effects on property prices. Landry et al. (2003) also found a negative effect, but their estimated effect was statistically insignificant. However, their contingent valuation study strongly suggests that beachgoers do not like beaches with seawalls. The existing literature suggests that our estimates might even be underestimated. Gopalakrishnan et al. (2011) correct for endogeneity bias and find that the beach width effect on property prices is almost five times larger. Furthermore, Gopalakrishnan et al. (2011) find that the long- term net value of coastal residential property can fall by as much as 52% when the erosion rate triples. To the extent that seawalls can potentially double or triple erosion rates on neighboring non-protected properties, our estimates are very much in line with those of Gopalakrishnan et al. (2011).

Our paper does not come with limitations. Aside from potential measurement problems in our price variable, our study may also be confounded by the potential immunity of the property market to climate risks, which may be coupled with households' potential downward bias against these risks. Potential buyers may also shy away from purchasing properties with visible protected measures, which may compel homeowners to opt not to install coastal defense measures. How these issues influence property housing, along with the determination of the most cost-effective approach to deal with increasing sea level rise, are some of the areas in which future research could explore on.

Despite the above methodological concerns, our paper represents a step forward in terms of measuring a previously unmeasured social cost (in California). Furthermore, the impact on house prices is only one of many ways that seawalls impose social costs and benefits. Recreational and existence values of beaches are also clearly important. Our paper finds that properties near beaches with significant seawall protection are selling for lower prices than those without. This information is easily observed, collected, and defensible. Even though we are still missing many of the social costs associated with seawalls, our hedonic estimates and implied welfare costs are so large that this raises serious questions about whether seawalls in California pass a simple cost-benefit analysis. The benefits to individual waterfront property owners would have to be extremely large to justify the social costs.

References

Arraiz, I., Drukker, D. M., Kelejian, H. H., and Prucha, I. R. (2010). A spatial cliff-ord-type model with heteroskedastic innovations: small and large sample results. *Journal of Regional Science*, 50(2):592–614.

Bin, O., Crawford, T. W., Kruse, J. B., and Landry, C. E. (2008a). Viewscapes and flood hazard: Coastal housing market response to amenities and risk. *Land Economics*, 84(3):434–448.

Bin, O., Kruse, J. B., and Landry, C. E. (2008b). Flood hazards, insurance rates, and amenities: Evidence from the coastal housing market. *Journal of Risk and Insurance*, 75(1):63–82.

Bin, O., Landry, C. E., Ellis, C. L., and Vogelsong, H. (2005). Some consumer surplus estimates for North Carolina beaches. *Marine Resource Economics*, 20(2):145–161.

Brown, G. M. and Pollakowski, H. O. (1977). Economic valuation of shoreline. *The Review of Economics and Statistics*, pages 272–278.

Caldwell, M. and Segall, C. H. (2007). No day at the beach: sea level rise, ecosystem loss, and public access along the California coast. *Ecology LQ*, 34:533.

Can, A. (1990). The measurement of neighborhood dynamics in urban house prices. *Economic Geography*, 66(3):254–272.

Cliff, A. D. and Ord, K. (1970). Spatial autocorrelation: a review of existing and new measures with applications. *Economic Geography*, 46(sup1):269–292.

Diamond, D. B. (1980). The relationship between amenities and urban land prices. *Land Economics*, 56(1):21–32.

Drukker, D. M., Peng, H., Prucha, I., and Raciborski, R. (2015). Sppack: Stata module for cross-section spatial-autoregressive models.

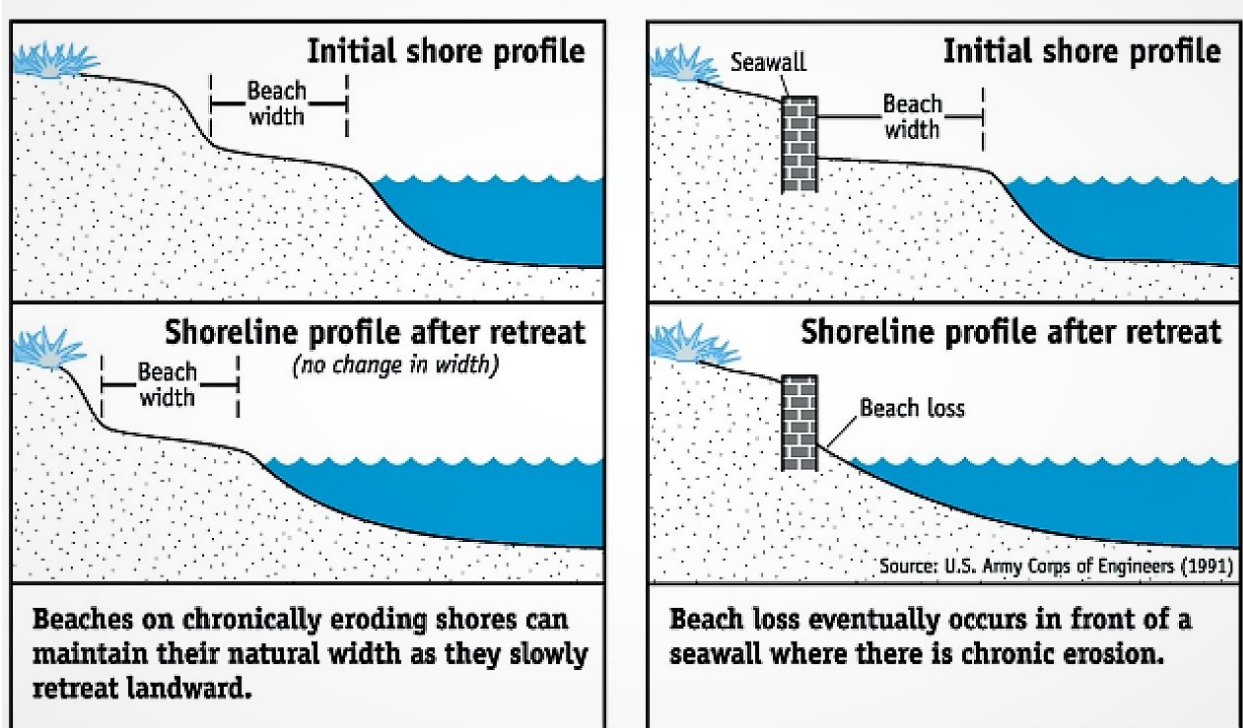Edwards, S. F. and Gable, F. J. (1991). Estimating the value of beach recreation from property val-

ues: An exploration with comparisons to nourishment costs. *Ocean and Shoreline Management*, 15(1):37–55.

Gopalakrishnan, S., Smith, M. D., Slott, J. M., and Murray, A. B. (2011). The value of disappearing beaches: a hedonic pricing model with endogenous beach width. *Journal of Environmental Economics and Management*, 61(3):297–310.

Griggs, G. B. (2005). The impacts of coastal armoring. *Shore and beach*, 73(1):13–22.

Griggs, G. B. (2009). The effects of armoring shorelines – the California experience. In *Puget Sound Shorelines and the Impacts of Armoring – Proceedings of a State of the Science Workshop*, pages 77–84.

Griggs, G. B. and Savoy, L. E. (1985). *Living with the California coast.* Durham, NC: Duke University Press.

Griggs, G. B. and Tait, J. F. (1988). The effects of coastal protection structures on beaches along northern Monterey Bay, California. *Journal of Coastal Research*, pages 93–111.

Hall, M. J., Young, R. S., Thieler, E. R., Priddy, R. D., and Pilkey Jr, O. H. (1990). horeline response to hurricane Hugo. *Journal of coastal research*, pages 211–221.

Heinz Center (2000). Evaluation of erosion hazards. The H. John Heinz III Center for Science. *Economics and the Environment, Washington.*

Kriesel, W. and Friedman, R. (2002). Coastal hazards and economic externality: implications for beach management policies in the American South East. Discussion paper, The H. John Heinz III Center for Science, Economics, and Management.

Kriesel, W. and Friedman, R. (2003). Coping with coastal erosion: evidence for community-wide impacts. *Shore and Beach*, 71(3):19–23.

Kriesel, W., Landry, C. E., and Keeler, A. (2005). Coastal erosion management from a community

economics perspective: The feasibility and efficiency of user fees. *Journal of Agricultural and Applied Economics*, 37(2):451–461.

Kriesel, W., Randall, A., and Lichtkoppler, F. (1993). Estimating the benefits of shore erosion protection in Ohio's Lake Erie housing market. *Water Resources Research*, 29(4):795–801.

Landry, C. E. and Hindsley, P. (2011). Valuing beach quality with hedonic property models. *Land Economics*, 87(1):92–108.

Landry, C. E., Keeler, A. G., and Kriesel, W. (2003). An economic evaluation of beach erosion management alternatives. *Marine Resource Economics*, 18(2):105–127.

Moran, P. A. (1950). Notes on continuous stochastic phenomena. *Biometrika*, 37(1/2):17–23.

Parsons, G. R. (1990). Hedonic prices and public goods: an argument for weighting locational attributes in hedonic regressions by lot size. *Journal of Urban Economics*, 27(3):308–321.

Parsons, G. R. and Powell, M. (2001). Measuring the cost of beach retreat. *Coastal Management*, 29(2):91–103.

Parsons, G. R. and Wu, Y. (1991). The opportunity cost of coastal land-use controls: An empirical analysis. *Land Economics*, 67(3):308–316.

Pendleton, L., Mohn, C., Vaughn, R. K., King, P., and Zoulas, J. G. (2012). Size matters: The economic value of beach erosion and nourishment in Southern California. *Contemporary Economic Policy*, 30(2):223–237.

Pisati, M. (2001). sg162: tools for spatial data analysis. *Stata Technical Bulletin*, 60:21–37.

Pompe, J. (2008). The effect of a gated community on property and beach amenity valuation. *Land Economics*, 84(3):423–433.

Pompe, J. J. (1999). Establishing fees for beach protection: paying for a public good. *Coastal Management*, 27(1):57–67.

Pompe, J. J. and Rinehart, J. R. (1995). Beach quality and the enhancement of recreational property values. *Journal of Leisure Research*, 27(2):143–154.

Taylor, L. O. (2003). The hedonic method. In Champ, P., Boyle, K., and T.C., B., editors, *A primer on nonmarket valuation. The Economics of non-market goods and resources*, volume 3, pages 331–393. Springer, Dordrecht.

Wu, S.-Y., Yarnal, B., and Fisher, A. (2002). Vulnerability of coastal communities to sea-level rise: a case study of Cape May County, New Jersey, USA. *Climate Research*, 22(3):255–270.

Appendix (For online publication)

Figure A.1: Shoreline Armoring and Beachloss

Table A.6: Description of Variables.

Variable	Description	Logged?	Source
House Size (square feet)	Size of the houses living area	Yes	Mail Survey & County Assessor
Lot Size (square feet)	Size of the land on the property parcel	Yes	Mail Survey & County Assessor
Bedrooms	Number of bedrooms in the home	Yes	Mail Survey & County Assessor
Geotime (years)	The expected number of years until the set-back distance is zero, given the propertys historical erosion rate	Yes	Heinz Center Field Measurement Team
Elevation (feet)	Elevation of first floor above base flood elevation	Yes	Heinz Center Field Measurement Team
Distance to Water (feet)	Distance to the ocean at the time the property was sold	Yes	Heinz Center Field Measurement Team
Beach Width (feet)	High tide beach width	Yes	Heinz Center Field Measurement Team
Time Since Transaction (years)	Number of years since property purchase	Yes	Mail Survey & County Assessor
Age of House (years)	Number of years since the house was constructed	Yes	Mail Survey & County Assessor
Fireplace (1-0)	1 if fireplace, = 0 otherwise	No	Mail Survey
Brick Exterior (1-0)	1 if brick or stone exterior finish	No	Mail Survey
Waterfront (1-0)	1 if parcel is waterfront, = 0 otherwise	No	Mail Survey & Heinz Field Measurement Team
Seawall (1-0)	1 if property is waterfront and its shore has hard stabilization (seawall)	No	Heinz Field Measurement Team
Percent Seawalls	The percent of waterfront properties within a community that were protected by seawalls	Yes	Heinz Field Measurement Team
Waterfront Nourishment (1-0)	1 if the property is waterfront and the beach has been nourished	No	Mail Survey & Heinz Field Measurement Team
Beach Nourishment (1-0)	1 if nearest beach has been nourished	No	Mail Survey & Heinz Field Measurement Team
Built Post-FIRM (1-0)	1 if house built after date of regular NFIP status, provides a measure of a given structures resistance to flood damage since it was constructed after the communitys acceptance into the Regular insurance program, when the tougher code was in effect	No	Mail Survey & Heinz Field Measurement Team
Property Price (2014 $)	Price the property was purchased for, converted to 2014 dollars	Yes	Mail Survey & County Assessor

Table A.7: Dependent Variable: Log(housing price/lot size).

Regression	All Variables			Reduced Model		
	San Diego	Santa Cruz	Combined	San Diego	Santa Cruz	Combined
Seawall (0-1)	0.144	-0.084	-0.122	0.282	-0.198	0.019
	(0.241)	(0.276)	(0.185)	(0.237)	(0.260)	(0.180)
Percent seawalled (%)	-0.031	-0.220**	-0.105**	0.006	-0.262***	-0.148***
	(0.047)	(0.088)	(0.052)	(0.059)	(0.072)	(0.051)
Waterfront (0-1)	0.050	0.628***	0.296*	-0.005	0.638***	0.351**
	(0.165)	(0.235)	(0.153)	(0.180)	(0.204)	(0.140)
Waterfront nourishment (1-0)	0.146	-0.589	-0.065	0.157	-0.328	-0.119
	(0.556)	(0.429)	(0.386)	(0.709)	(0.398)	(0.350)
Beach nourishment (1-0)	0.059	0.697*	0.263	-0.212	0.766***	0.458*
	(0.447)	(0.395)	(0.315)	(0.389)	(0.271)	(0.234)
House size (sq. ft.)	0.369***	-0.516**	0.170			
	(0.107)	(0.211)	(0.107)			
Years since last transaction	-0.404***	-0.346***	-0.386***			
	(0.063)	(0.086)	(0.054)			
Years since built	-0.310**	-0.919***	-0.549***			
	(0.152)	(0.204)	(0.126)			
Fireplace (0-1)	0.171	0.130	0.113			
	(0.119)	(0.187)	(0.112)			
Bedrooms	-0.250**	-0.105	-0.309**			
	(0.123)	(0.199)	(0.122)			
Geotime	-0.218	-0.013	-0.197*			
	(0.382)	(0.116)	(0.105)			
Geotime squared	0.010	-0.003	0.011			
	(0.024)	(0.010)	(0.007)			
Elevation (ft.)	-0.005	0.080***	0.046**			
	(0.030)	(0.030)	(0.023)			
Distance to water (ft.)	-0.011	-0.129	-0.048			
	(0.121)	(0.078)	(0.074)			
Beach width	-0.010	-0.003	-0.018			
	(0.033)	(0.037)	(0.026)			
Built Post-FIRM	-0.883**	-1.338**	-1.165***			
	(0.412)	(0.565)	(0.316)			
San Diego (1-0)			0.085			-0.371***
			(0.138)			(0.104)
Constant	5.046***	13.263***	7.393***	4.236***	5.278***	5.012***
	(1.570)	(1.732)	(0.910)	(0.203)	(0.258)	(0.187)
Observations	130	117	247	132	151	283
R-sq.	0.433	0.568	0.460	0.023	0.191	0.126
R-sq. (adj.)	0.352	0.498	0.420	-0.016	0.163	0.107

Note: Robust standard errors are in parentheses. *,**,*** denote statistical significance at 10, 5, and 1 percent, respectively.

Table A.8: Tests for Spatial Dependence.

	San Diego	Santa Cruz
Spatial Error (SEM)		
Moran's I	0.100	1.265
Lagrange Multiplier	0.276	0.026
Robust Lagrange Multiplier	0.230	1.289
Spatial Lag (SLM)		
Lagrange Multiplier	0.080	0.947
Robust Lagrange Multiplier	0.033	2.210

Note: The test are performed in an OLS framework using our base model
(equation 2) and following the procedure of Pisati (2001). *,**,*** denote
statistical significance at 10, 5, and 1 percent, respectively.

Calculating Geotime Variable

Step 1: Computing Erosion Rates

In the field measurement survey, two distance variables were collected: Distance to Current ERF (dist-cerf) and Distance to 60-year Projected ERF (dist-perf). ERF is the Erosion Reference Feature, typically the top edge of a bluff, dune, vegetation line or beach. We denote the erosion rate as realER.

I. If dist-cerf > dist-perf, then erosion rate is calculated by taking the difference and dividing by 60. Thus a positive erosion rate will be generated, with units of feet reduced per year.

II. If dist-cerf $\leq$ dist-perf, and the shore is neither protected by armoring nor nourishment, a rate of 0.01 is assigned to the property.

III. If dist-cerf $\leq$ dist-perf, and the shore is protected, we use weighted average erosion rates, which are generated from the sample of unprotected coastal properties within each community.

Step 2: Compute Distance to Water

To compute Distance to Water at the time that the property was sold, we need to add the estimated distance that had been erased between that year and when the field survey was conducted. Distance to Water = dist-cerf + (Time Since Transaction * realER)

The value of Distance to Water is truncated above one, such that its natural logarithm will be non-negative.

Step 3: Compute Geotime

Geological time, Geotime, is defined as the number of years until the propertys Distance to Water is reduced to zero by erosion. The formula to compute Geotime is as follows.

Geotime = Distance to Water/realER