Decadal-scale coastal cliff retreat in southern and central California

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A B S T R A C T

Airborne LiDAR data collected in 1998 and 2009–2010 were used to measure coastal cliff erosion and retreat between the Mexico/California border and Bodega Head, California. Cliff erosion was detected along 44% of the 595 km of shoreline evaluated, while the remaining cliffs were relatively stable. The mean cliff top retreat rate was 0.12 m/yr, while mean retreat averaged over the entire cliff face was 0.04 m/yr. The maximum cliff top and face retreat rates were 4.2 and 3.6 m/yr, respectively. Historical (~1930s to 1998) and recent retreat rates were significantly inversely correlated for areas with large historical or recent cliff retreat, such that locations with elevated historical retreat had low levels of recent retreat and locations with elevated recent retreat were preceded by low rates of historical retreat. The strength of this inverse correlation increased with cliff change magnitudes up to $r^2$ of 0.91 for cliff top retreat rates $>2.9$ m/yr. Mean recent retreat rates were 52-83% lower than mean historical retreat rates. Although beaches can protect cliffs against wave-driven erosion, cliffs fronted by beaches retreated 49% more than cliffs without beaches. On average, unarmored cliff faces retreated 0.05 m/yr between 1998 and 2009–2010, about three times faster than artificially armored cliffs. Alongshore metrics of wave-cliff impact, precipitation, and cliff hardness were generally not well correlated with recent cliff changes. A cliff hazard metric is used to detect cliff steepening and areas prone to future cliff top failures.

1. Introduction

Coastal cliffs comprise a high proportion of the world’s coasts (Emery and Kuhn, 1982), where almost one quarter of the global population resides (Small and Nicholls, 2003). Retreating coastal cliffs currently cause numerous problems for coastal populations and managers. Sea-level rise is expected to cause increased coastal cliff erosion rates for many areas (Sunamura, 1988; Bray and Hooke, 1997; Dickson et al., 2007; Nicholls et al., 2007; Trenhaile, 2010, 2014). Studies of coastal cliff erosion and retreat have increased in recent years (Naylor et al., 2010), highlighting a growing interest and need to study rock cliffs. Yet our understanding of coastal cliff processes and behavior is complicated by the wide array of erosional processes that occur (Trenhaile, 1987; Sunamura, 1992; Rosser et al., 2007; Kennedy et al., 2011), variable temporal changes and processes (Cambers, 1976; Dornbusch et al., 2008; Lee, 2008), geomorphic feedbacks (Sunamura, 1976; Kline et al., 2014; Young, 2015), and highly variable geologic, oceanographic, and climatic settings.

Actively eroding coastal cliffs comprise the majority of the California coast (Fig. 1) and threaten development throughout the State, including highways, railways, wastewater, oil, natural gas and nuclear facilities, universities, military bases, and numerous state beaches and parks in addition to homes and businesses. Episodic cliff failures have caused human injury and several deaths in recent years (e.g. Perry, 2000; Gross and Davis, 2008; Evans, 2015). Seawalls and rock armoring are increasingly used to prevent erosion, but eroding coarse-grained coastal cliffs can be an important source of sediment to beaches (Young and Ashford, 2006; Brooks and Spencer, 2010; Young et al., 2010a; Mushkin et al., 2016), which are important cultural and economic resources that generate billions of dollars annually in California. These problems will worsen as sea levels and coastal populations continue to increase, and effectively managing California’s changing coast will become increasingly challenging.

Coastal cliff erosion is broadly attributed to marine and subaerial (including subsurface) erosion mechanisms (Trenhaile, 1987; Sunamura, 1992). Subaerial mechanisms (e.g. groundwater processes, rilling, slope wash) act over the entire cliff face, and beneath the surface. Rainfall has been empirically linked to inland landsliding (Caine, 1980), where marine processes are not active, and serves as an indicator of subaerial forcing. Young et al. (2009b) found a high correlation between the timing of rainfall and coastal cliff erosion in southern California. Brooks et al. (2012) discussed the importance of the sequence of rain events on sub-surface pore water pressures and cliff stability. Marine processes (e.g. wave-driven impact pressures and abrasion) act directly only at the cliff base, and only when tides and other water level fluctuations allow waves to reach the cliff (Sunamura, 1992; Rosser et al., 2013; Vann Jones et al., 2015; Young et al., 2016). While marine and subaerial processes drive cliff erosion, geologic conditions dictate cliff resistance and control the seacliff failure mode. The relative importance of marine and subaerial processes can be temporally dynamic, especially during extreme weather.
processes varies in space and time, and observations of cliff erosion and forcing (e.g., ocean waves and rain) are needed to establish these relationships. Brooks et al. (2012) established both marine and subaerial process thresholds associated with high magnitude cliff retreat events for soft cliffs on the Suffolk coast, U.K. and concluded the driving mechanisms vary through time.

Historical cliff retreat has been measured with historical maps and aerial photographs in a variety of locations worldwide, including Portugal (Marques, 2008), the United Kingdom (May and Heeps, 1985; Brooks and Spencer, 2010), Italy (Budetta et al., 2000), Australia (Bezore et al., 2016), Canada (Lantuit and Pollard, 2008), and India (Sajinkumar et al., 2017). LiDAR is increasingly used to measure more recent cliff retreat and erosion and has been applied in the United Kingdom (Adams and Chandler, 2002; Rosser et al., 2005; Earlie et al., 2015), Portugal (Nunes et al., 2011), the United States (e.g., Sallenger et al., 2002; Young and Ashford, 2006), Canada (Obu et al., 2016), Israel (Katz and Mushkin, 2013) and France (Letortu et al., 2015). Other methods to measure cliff retreat and erosion include using fixed markers in the United Kingdom (Williams and Davies, 1987), field surveys in Australia (Gill, 1973), micro-erosion meters (MEM) in New Zealand (Stephenson and Kirk, 1996) and measuring shore platform widths in New Zealand (de Lange and Moon, 2005).

Coastal cliff retreat studies in California date back at least to 1932 (Vaughan, 1932). Since then, numerous studies of California cliff retreat have been conducted using a variety of measurement techniques ranging from dated inscriptions (Emery and Kuhn, 1980) to terrestrial LiDAR (Collins and Sitar, 2008). These studies are often site-specific or local in scale, but Griggs et al. (2005) and Darl (2005) provide statewide compilations of many of these studies, and U.S. Army Corps of Engineers (1971) provides a qualitative statewide erosion assessment. Hapke et al. (2009) conducted the most recent systematic study and mapped retreat along 350 km of California cliffs using T-sheets from the 1920s and 1930s and airborne LiDAR data collected in 1998 and 2002. Using shore-normal transects spaced 20 m alongshore, Hapke et al. (2009) calculated mean and maximum cliff top retreat rates of 0.3 m/yr and 3.1 m/yr, respectively, with estimated errors of 0.2 m/yr. The highest retreat rates were in central and northern California, associated with large slumps and deep-seated coastal landslide activity, often triggered by elevated rainfall and associated increased pore water pressures (Thomas and Loague, 2014; Young, 2015).

Advances in cliff mapping with airborne LiDAR surveys (Matsumoto et al., 2017) provide unprecedented detail of volumetric and cliff top change over large areas, but existing studies in California have limited spatial extent (i.e., Young et al., 2009a, 2010b, 2011). This study provides the first large-scale assessment of coastal cliff erosion and retreat in California using high-resolution airborne LiDAR data collected in 1998 and 2009–2010. Both cliff top retreat and the average retreat of over the cliff face are quantified and used to explore changes on different parts of the cliff profile. The results are compared with historical retreat rates, alongshore metrics of erosion forcing mechanisms, and rock strength. A cliff hazard metric described by Young et al. (2009a) is used to locate areas prone to future cliff top failures.

2. Study site

2.1. Geologic setting

The study site extends 1100 km from the Mexico/California border to Bodega Head, California (Fig. 1). The California coast is tectonically active and contains numerous fault zones, most notably the San Andreas Fault dividing the North American and Pacific plates. Past tectonic processes produced several coastal mountain ranges and a series of uplifted marine terraces along much of the coastline. The majority of coastal...
cliffs are low relief cliffs cut into the uplifted marine terraces while the rest are high relief cliffs and coastal mountains (Griggs et al., 2005).

Cliffs cut into the marine terraces are generally composite cliffs composed of two geologic units: a more resistant lithified Cenozoic mudstone, shale, sandstone and siltstone, and an upper unit of weakly lithified Quaternary terrace deposits (Griggs et al., 2005). Low-relief cliffs composed of alluvium or terrace deposits also exist where the lower unit is locally absent. The cliffs are generally 10–30 m in height, but exceed 100 m in some areas. Cliff heights remain generally constant as the cliff recedes because most cliff tops are relatively flat terrace features. The cliffs are typically fronted by a wave-cut platform usually covered by a veneer of beach sand and sometimes cobble. The beaches, which act as a buffer to direct wave-driven cliff erosion, are often narrow and occasionally stripped of sediment during large winter storms. Cliff profiles differ alongshore and their geometries are related to the relative importance of marine and subaerial processes, cliff composition, and phases of cliff profile evolution (Emery and Kuhn, 1982).

2.2. Oceanography

The California coast is exposed to waves generated by local winds and distant storms in both hemispheres (Flick, 1994). During winter, swell from the North Pacific and Gulf of Alaska is most energetic, whereas swell from the South Pacific dominates in summer. Waves reaching the southern California coast undergo a complex transformation, and shadows of the Channel Islands create strong alongshore variations in the southern California coast undergo a complex transformation, and as swell from the South Pacific and Monterey areas. Strong El Niño events are usually dri.edu), but is locally higher along the Big Sur coast and lower in the 257 mm in San Diego to 1032 mm near Bodega Head (www.wrcc.e,Et al., 1987; Haston and Michaelsen, 1994). Annual coastal precipitation generally increases northward with mean annual precipitation ranging from 257 mm in San Diego to 1032 mm near Bodega Head (www.wrcc.et al., 1987; Haston and Michaelsen, 1994). Annual coastal precipitation generally increases northward with mean annual precipitation ranging from 257 mm in San Diego to 1032 mm near Bodega Head (www.wrcc.edu). Prolonged drought, very wet years, and multi-decadal scale precipitation cycles can cause high variation in annual rainfall (Michaelsen et al., 1987; Haston and Michaelsen, 1994). Annual coastal precipitation generally increases northward with mean annual precipitation ranging from 257 mm in San Diego to 1032 mm near Bodega Head (www.wrcc.dri.edu), but is locally higher along the Big Sur coast and lower in the San Francisco and Monterey areas. Strong El Niño events are usually associated with elevated winter precipitation, wave heights, and sea levels, causing increased coastal erosion, flooding and damage (Flick, 1998; Storlazzi and Griggs, 2000; Storlazzi et al., 2000; Barnard et al., 2017). Annual precipitation during the study was near average except during the winter of 2004–2005 when some regions received more than twice the annual mean precipitation.

3. Methods

3.1. LiDAR data

Airborne LiDAR datasets collected in 1998 and 2009–2010 (coast.noaa.gov/digitalcoast/; Table 1) provide regional coverage from the California-Mexico border to Bodega Head, with some gaps, notably along the Big Sur Coast (Fig. 1). A 2002 LiDAR dataset provides additional spatial coverage but was not used here because of low point density (probably from fewer aircraft flight passes) and data gaps. However, a cliff top line digitized by Hapke and Reid (2007) from the 2002 LiDAR was used in some localized areas. The overall coverage for this study includes 595 km of cliffs.

The LiDAR survey contractors estimated vertical root mean square error of the 1998 and 2009–2010 point data at 0.15 m, with horizontal accuracies of 0.5–0.80 m. LiDAR point data were processed into 1-m resolution digital elevation models using the last return (if multiple returns were available) and a natural neighbors technique. Typical point density of the 1998 and 2009–2010 datasets are 0.5 and 1.5 points/m², respectively.


Digital change grids, estimated by differencing successive digital elevation models created using these LiDAR datasets, show both negative (erosion) and positive (accretion, talus deposits) changes. Sources of digital change grid error include the basic LiDAR observations, spatial interpolation, and vegetation. The vertical root mean square difference (RMSZ) between surveys was estimated at 0.36 m using 14 control areas spread throughout the study area. Control areas representative of coastal cliff topography consisted of inland hillsides and slopes assumed to experience no change during the study period. RMSZ was calculated for each control area using the digital change grid raster values.

Elevation changes can indicate landslide motion, land erosion, talus deposition, topographic beach changes, and anthropogenic changes. The detected changes were programatically filtered to remove noise and erroneous data. To do so, first all grid cells with a vertical change

<table>
<thead>
<tr>
<th>South End</th>
<th>North End</th>
<th>Survey date</th>
<th>Analysis time span (years, 1998 to 2009–2010)*</th>
</tr>
</thead>
<tbody>
<tr>
<td>Border</td>
<td>La Jolla Shores</td>
<td>November 2009</td>
<td>11.0</td>
</tr>
<tr>
<td>La Jolla Shores</td>
<td>Dana Point Harbor</td>
<td>October 2009</td>
<td>11.5</td>
</tr>
<tr>
<td>Dana Point Harbor</td>
<td>Seal Beach</td>
<td>October 2009</td>
<td>11.5</td>
</tr>
<tr>
<td>Seal Beach</td>
<td>Point Dune</td>
<td>October 2009</td>
<td>11.6</td>
</tr>
<tr>
<td>Point Dune</td>
<td>Carpenteria</td>
<td>November 2009</td>
<td>11.6</td>
</tr>
<tr>
<td>Carpenteria</td>
<td>El Capitan</td>
<td>November 2009</td>
<td>11.6</td>
</tr>
<tr>
<td>El Capitan</td>
<td>Point Conception</td>
<td>November 2009</td>
<td>11.6</td>
</tr>
<tr>
<td>Point Conception</td>
<td>Oceano</td>
<td>November 2009</td>
<td>11.6</td>
</tr>
<tr>
<td>Oceano</td>
<td>Harmony</td>
<td>November 2009, May 2010</td>
<td>11.0</td>
</tr>
<tr>
<td>Harmony</td>
<td>Ragged Point</td>
<td>November 2009</td>
<td>11.6</td>
</tr>
<tr>
<td>Ragged Point</td>
<td>Point Sur</td>
<td>November 2009, May 2010</td>
<td>11.7</td>
</tr>
<tr>
<td>Point Sur</td>
<td>Moss Landing</td>
<td>May 2010, June 2010, October 2010</td>
<td>12.1</td>
</tr>
<tr>
<td>Moss Landing</td>
<td>Ano Nuevo</td>
<td>June 2010, September 2010, October 2010</td>
<td>12.2</td>
</tr>
<tr>
<td>Ano Nuevo</td>
<td>Pacifica</td>
<td>June 2010, September 2010, November 2010</td>
<td>12.2</td>
</tr>
<tr>
<td>Pacifica</td>
<td>Golden Gate</td>
<td>June 2010, October 2010, November 2010</td>
<td>12.2</td>
</tr>
<tr>
<td>Golden Gate</td>
<td>Tomales Bay</td>
<td>September 2010, November 2010</td>
<td>12.4</td>
</tr>
<tr>
<td>Tomales Bay</td>
<td>Sea Ranch</td>
<td>September 2010, November 2010</td>
<td>12.4</td>
</tr>
</tbody>
</table>

* Longest date range.
of <1 m (about 3 × RMSZ) were omitted. Next, a minimum topographic footprint was imposed, requiring >10 connected cells of positive or negative change. Finally, the filtered digital change data were checked visually and edited manually. Changes related to beaches, dunes, construction, and those areas inland of the coastal road were identified using aerial photographs and digital elevation model hillshades, and removed.

Changes were separated into negative (i.e. cliff erosion) and positive (i.e. talus deposits) volumetric changes and then evaluated in 5 m wide (in the alongshore direction) compartments. Dividing the volumetric compartment changes by the cliff height and compartment width (5 m) yielded bulk negative and positive cliff face changes, equivalent to average cliff retreat/advance over the cliff face (Fig. 2). Cliff/coastal slope heights were obtained from the digital elevation model.

The calculated volume changes underestimate the actual changes because only relatively large volume and large footprint slides are detected. Thus, smaller topographic changes (such as a smaller individual rockfalls or localized surficial erosion) were not detected with the present methods. In a few locations such as Palos Verdes and Big Sur, the LiDAR swath did not fully cover the landward extent of coastal change, also causing an under-estimation of coastal change volumes.


Cliff top changes were measured using an existing 1998 (or 2002 in some locations) cliff top edge line from Hapke and Reid (2007) and a 2009–2010 cliff top edge line digitized manually for this study. The cliff top edge location was defined as the slope break between the cliff face and the cliff top. Areas were excluded where a definitive edge could not be identified. Cliff top retreat was measured at 5 m intervals alongshore using shore-normal transects. Negative cliff top retreat indicates landward cliff top movement (erosion). Measurements indicating seaward cliff top movement were assumed to result from data processing artifacts, and set to ‘no change’. In some localized areas, cliff top retreat measurements were removed because of variation in the identified cliff top edge feature between this study and Hapke and Reid (2007). Cliff edge uncertainties of 1.3 and 1.1 m for the 1998 and 2009–2010 datasets, respectively, were estimated as the sum in quadrature of the horizontal LiDAR accuracy (0.5 and 0.8 m) and digitizing error estimated at 1 m. The annualized error for cliff top retreat rates was estimated as the sum in quadrature of the two cliff edge uncertainties divided by the time span (11.6–12.4 years) at 0.14–0.15 cm/yr.


Historical cliff edge lines from 1929 to 1934 and 1998/2002 (previously analyzed by Hapke and Reid, 2007) were reanalyzed at 5 m alongshore resolution for spatial consistency with the present analysis. Measurements indicating seaward cliff top movement were set to ‘no change’. The error associated with these cliff retreat rates is estimated at 0.20 m/yr (Hapke and Reid, 2007).

3.5. Waves and total water level

Hourly tide levels were obtained from the La Jolla, Los Angeles, Santa Monica, Santa Barbara, Port San Luis, Monterey, and Point Reyes tide gauges (tidesandcurrents.noaa.gov). A wave buoy network (CDIP, cdip.ucsd.edu) was used to estimate hourly wave conditions at 8707 virtual buoys located in 10 m water depth spaced at 100 m intervals alongshore. The effects of complex bathymetry, and of varying beach orientation and wave exposure, were simulated with a spectral refraction wave model initialized with offshore buoy data (O’Reilly and Guza, 1991, 1998). The model has been used throughout California and validated extensively (O’Reilly et al., 2016). Time periods without data at all virtual buoys were removed, resulting in 46,258 h between November 2003 and November 2009 for wave data analysis.

The total water level is the sum of tides and the vertical height of wave run-up (Shih et al., 1994; Kirk et al., 2000; Ruggiero et al., 2001). Time series of hourly total water level at the cliff base provide basic estimates of wave impact duration and a proxy for marine forcing. The vertical height of wave run-up (%R2) was approximated as the level exceeded by 2% of wave uprushes (Stockdon et al., 2006), using the formula:

\[ R2\% = 0.043(H_{w,a})^{0.5} \]

where the deep water wave height (Hw) used in the run-up equation was calculated by backing out nearshore wave height to deep water by reverse shoaling using linear wave theory, while the deep water wavelength (Lw) was calculated using the linear dispersion relationship (Dean and Dalrymple, 1991). Hourly total water level at each cliff location was interpolated with R2% and tide gauge data and used to calculate wave–cliff impact metrics, defined as the number of hours total water levels exceeded 2.5 and 3 m in elevation (NAVD88).

<table>
<thead>
<tr>
<th>Cliff Erosion (m3/m)</th>
<th>a</th>
<th>b</th>
<th>c</th>
<th>d</th>
<th>e</th>
</tr>
</thead>
<tbody>
<tr>
<td>Volume (-)</td>
<td>-2.0</td>
<td>-2.0</td>
<td>-2.0</td>
<td>-2.0</td>
<td>-2.0</td>
</tr>
<tr>
<td>Volume (+)</td>
<td>0.0</td>
<td>0.0</td>
<td>0.0</td>
<td>+2.0</td>
<td>+2.0</td>
</tr>
<tr>
<td>Volume (Net)</td>
<td>-2.0</td>
<td>-2.0</td>
<td>-2.0</td>
<td>0.0</td>
<td>0.0</td>
</tr>
</tbody>
</table>

Fig. 2. Interpretations of idealized cliff changes using the different retreat and erosion estimates.
3.6. Coastal armoring and setting

Coastal armoring present in 2010 was mapped by updating and amending a 2004 database of coastal armoring (Dare, 2005) using 2010 oblique coastal photographs (California Coastal Records Project). The general coastal setting (rocky coast, sandy beach, cliff fronted by beach, etc.) was categorized and mapped alongshore using maps from Griggs et al. (2005) and 2010 oblique photographs (California Coastal Records Project). An accurate time series of beach width could not be established for the study area because of limited data availability and variable survey dates (Table 1) combined with large seasonal beach fluctuations, sometimes >50 m (Doria, 2016).

3.7. Rock strength

Nine hundred and thirty three in situ cliff rock strength measurements were made with Proceq Schmidt Hammers Type L and N at the cliff base using the ASTM method (ASTM, 2013). The ASTM method calculates the rebound value as the mean of 10 readings with outliers removed. Type L and N hammers were compared using 20 in situ measurements on a variety of rock types and combined with measurements from Kennedy and Dickson (2006) for calibration. Type L and N measurements were highly correlated ($r^2 = 0.98$), and Type N hammer measurements were converted to Type L for consistency. Schmidt hammer measurements were typically spaced 50–100 m alongshore in the sections evaluated. For each cliff compartment rebound measurements located within 100 m were used to evaluate rock rebound statistics including the minimum, maximum, mean, and non-zero mean. Similar rock rebound metrics were also evaluated using rebound measurements within 500 and 1000 m of each cliff compartment. Site access logistics limited most rock strength measurements to cliffs fronted by beaches.

3.8. Precipitation

Total precipitation during the study period was estimated by interpolating daily precipitation records from 18 coastal sites distributed over the study area to each 5 m alongshore compartment (Fig. 1; www.wrcc.dri.edu/).

![Fig. 3. Alongshore comparison of (a) coastal setting and classification, (b) cliff or coastal slope height, (c) total water level index (hours above 2.5 m NAVD88), (d) interpolated study period total precipitation, (e) Schmidt hammer rebound metrics (within 1 km of compartment location) and (f) coastline reference map with hashes at 200 km intervals. Lower rebound metrics correspond to weaker rocks.](image-url)
Table 2
Coastal setting, cliff change, and cliff top hazard statistics by county.

<table>
<thead>
<tr>
<th>County</th>
<th>Length in study area (km)</th>
<th>Classification (%)</th>
<th>Mean</th>
<th>Mean</th>
<th>Mean</th>
<th>Mean</th>
<th>Mean</th>
<th>Mean</th>
<th>Mean</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td></td>
<td>Armored Beach Cliff &amp; Rocky Bay/Harbor</td>
<td>wave impact index</td>
<td>rock rebound value</td>
<td>recent cliff top retreat (m/yr)</td>
<td>recent cliff face retreat (m/yr)</td>
<td>cliff top evaluated (km)</td>
<td>cliff face evaluated (km)</td>
<td>cliff top hazard index</td>
</tr>
<tr>
<td>Sonoma</td>
<td>7</td>
<td>2</td>
<td>42</td>
<td>25</td>
<td>26</td>
<td>7</td>
<td>2538</td>
<td>NaN</td>
<td>−0.32</td>
</tr>
<tr>
<td>Marin</td>
<td>111</td>
<td>3</td>
<td>21</td>
<td>42</td>
<td>33</td>
<td>3</td>
<td>2872</td>
<td>28</td>
<td>−0.21</td>
</tr>
<tr>
<td>San Francisco</td>
<td>19</td>
<td>23</td>
<td>26</td>
<td>23</td>
<td>23</td>
<td>28</td>
<td>2375</td>
<td>32</td>
<td>−1.08</td>
</tr>
<tr>
<td>San Mateo</td>
<td>82</td>
<td>10</td>
<td>13</td>
<td>52</td>
<td>32</td>
<td>3</td>
<td>3951</td>
<td>27</td>
<td>−0.16</td>
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<tr>
<td>Santa Cruz</td>
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<td>24</td>
<td>18</td>
<td>49</td>
<td>33</td>
<td>0</td>
<td>1854</td>
<td>34</td>
<td>−0.09</td>
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<tr>
<td>Monterey</td>
<td>64</td>
<td>4</td>
<td>20</td>
<td>13</td>
<td>66</td>
<td>1</td>
<td>2547</td>
<td>NaN</td>
<td>−0.13</td>
</tr>
<tr>
<td>San Luis Obispo</td>
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<td>8</td>
<td>24</td>
<td>21</td>
<td>53</td>
<td>1</td>
<td>1963</td>
<td>NaN</td>
<td>−0.04</td>
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<tr>
<td>Santa Barbara</td>
<td>176</td>
<td>12</td>
<td>22</td>
<td>70</td>
<td>8</td>
<td>0</td>
<td>1317</td>
<td>37</td>
<td>−0.11</td>
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<tr>
<td>Ventura</td>
<td>67</td>
<td>55</td>
<td>52</td>
<td>31</td>
<td>15</td>
<td>2</td>
<td>399</td>
<td>NaN</td>
<td>−0.09</td>
</tr>
<tr>
<td>Los Angeles</td>
<td>117</td>
<td>27</td>
<td>23</td>
<td>51</td>
<td>13</td>
<td>14</td>
<td>325</td>
<td>38</td>
<td>−0.09</td>
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<tr>
<td>Orange</td>
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<td>42</td>
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<td>35</td>
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<td>4</td>
<td>380</td>
<td>37</td>
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</tr>
<tr>
<td>San Diego</td>
<td>120</td>
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<td>40</td>
<td>45</td>
<td>12</td>
<td>3</td>
<td>558</td>
<td>16</td>
<td>−0.14</td>
</tr>
<tr>
<td>All</td>
<td>1133</td>
<td>18</td>
<td>27</td>
<td>40</td>
<td>30</td>
<td>3</td>
<td>1683</td>
<td>18</td>
<td>−0.12</td>
</tr>
</tbody>
</table>

a Compartment means for wave and rock metrics shown in Fig. 2.

3.9. Cliff top hazard index

Over long periods, cliff face and cliff top retreat measurements will converge. However, over shorter time periods, these measures can differ substantially and provide information on geomorphic change and cliff stability. Cliff top retreat reduces the overall cliff slope, while cliff base and cliff face erosion (not concentrated at the cliff top) cause slope steepening, thus reducing overall cliff stability. Cliff top retreat reduces the overall cliff slope, while cliff base retreat exceeds cliff top retreat, the cliff becomes more unstable, and vice versa. A cliff top hazard index, defined here as the recent cliff top retreat rate (positive values set to zero) minus the recent cliff face net retreat, increases with overall cliff steepening. Positive hazard values indicate the cliff face retreat rates exceed the cliff top retreat rates, suggesting a higher potential for future cliff top failure.

4. Results

4.1. Coastal setting

Seventy percent of the studied shoreline contains coastal cliffs, with 57% fronted by beaches (Fig. 3a, Table 2). Beaches without cliffs occupy 27% of the shoreline, with the remaining 3% of the coastline consisting of harbors and waterways. Total water level index metrics were consistently higher north of Point Conception (Fig. 3c). Eighteen percent of alongshore compartments contained some level of coastal armoring (Fig. 3a). Armoring is more prevalent in southern California probably because of higher density coastal populations and development, despite lower nearshore wave energies. Rock rebound values were relatively low in the San Diego area (Table 2), but generally varied widely in the sampled areas (Fig. 3e).

4.2. Cliffs changes

Recent (1998 to 2009–2010) cliff changes > 10 m³ were detected in 45% of the 5 m alongshore compartments (267 of 595 km; Table 3, Fig. 4d). Recent net volumetric cliff change rates ranged from −527 (erosion) to 128 (accretion) m³/m/yr with a mean of −1.78 m³/m/yr (Table 3). Recent net cliff face retreat ranged from −3.8 (landward) to 0.67 (seaward) m/yr with a mean of −0.042 m/yr (Table 3, Fig. 4c). Recent cliff top retreat rates ranged from −4.2 to 0.0 m/yr, with a mean of −0.12 m/yr (Table 3, Fig. 4b) for the 236 km evaluated. In comparison, historical (1929–1934 to 1998/2002) cliff top retreat rates ranged from −3.1 to 0.0 m/yr, with a mean of −0.25 m/yr (Table 3, Fig. 4a) for the 283 km evaluated. Across the 169 km of cliffs where both historical and recent cliff top rates exist, the mean historical and recent cliff top retreat rates were −0.22 and −0.12 m/yr, respectively.

Numerous cliff erosion and retreat hot spots were detected throughout the study area (Fig. 5). Many of the erosion hot spots were related to deep-seated and/or complex coastal slumps and/or complex coastal landslides such as in Daly City,
Portuguese Bend, and San Onofre, consistent with previous studies (Hapke et al., 2009; Young et al., 2009b; Young, 2015).

4.3. Cliff top hazard index

The cliff top hazard index ranged from $-3.7$ to $1.7$ (Fig. 4e) with a mean of $-0.07$, indicating more areas experienced cliff flattening as opposed to cliff steepening. However, 33% of compartments with both cliff top retreat and cliff face retreat observations had a positive hazard index. Locations with large cliff top hazard values ($>1$) include San Onofre State Beach, Big Sur, Martin’s Beach, Daly City, and near Double Point in Point Reyes National Seashore. San Luis Obispo County had the highest overall mean hazard index but the mean index of cliff steepening locations (positive hazard values) was relatively low (Table 2). Marin, San Francisco, and San Mateo Counties, all in the northern part of the study area, had the highest mean index of cliff steepening locations. Marin, San Luis Obispo, and Santa Barbara Counties had the highest percentages (42–62%) of cliff steepening locations (Table 2).

Additional future surveys are needed to test the applicability of this experimental hazard index.

5. Discussion

5.1. Influence of coastal setting

Correlations between total water level metrics and recent cliff changes were not significant at spatial scales between 5 m–0.5 km (averaging alongshore), but became statistically significant ($p < 0.01$) at 0.5 km scale, and increased to a maximum of $-0.15$ ($r^2$) at 15 km scales. At larger scales, the correlations became unstable with high $r^2$ fluctuations. Although not available for this study, incorporating time series data of beach widths and elevations to assess wave-cliff impacts may improve correlations between marine forcing and cliff retreat. Neither precipitation nor rock rebound metrics were well correlated with cliff retreat at any spatial scales. This could be because of the relatively low sampling resolution for rainfall and rock strength compared with cliff changes, or that other local parameters such as sea spray,
wind, and beach elevations are influencing cliff retreat. For example, rainfall runoff directed onto specific cliff areas can create erosion hot spots unrelated to cliff hardness or regional rainfall quantities. Additionally, the decadal time span averages out high magnitude rain and wave events causing the cliff erosion. The processes driving cliff erosion vary in space and time, and alongshore differences in cliff stage profile development and the stochastic nature of cliff failures probably also limited cliff erosion correlations. As identified in previous work (e.g. Young et al., 2009b; Young, 2015), rainfall may be better correlated with cliff erosion when compared in time series rather than the spatial comparison used here because time series analysis of a particular cliff section reduces cliff setting variables (such as geologic conditions) that are difficult to quantify.

Mean cliff face and cliff top retreat rates in central California were 86% and 30% larger, respectively, compared to southern California. Similarly, mean metrics of wave-impact and precipitation were also higher in central California, suggesting possible relationships at these regional scales; however additional data are needed to test this relationship statistically.

On average, unarmored cliffs retreated ($-0.054$ m/yr cliff face retreat) about 3 times more than armored cliffs ($-0.019$ m/yr, Fig. 6a). Cliffs fronted by beaches retreated 49% more ($-0.061$ m/yr cliff face retreat) than those without beaches ($-0.041$ m/yr, Fig. 6a). This observation is counter-intuitive because beaches protect cliffs from wave erosion (Jones and Williams, 1991; Lee, 2008), but is consistent with previous studies (Robinson, 1977). This observation suggests beaches in the study area tend to form at locations with relatively weak cliffs and where sufficient sand supply exists. It also highlights the possible role of beach sediment as an abrasive that may accelerate cliff retreat (Sunamura, 1982, 1992; Kline et al., 2014). Additional rock

Fig. 5. Examples of significant coastal cliff changes at (a) Point Reyes National Seashore, (b) Martin’s Beach, (c) Daly City and Pacifica, (d) Portuguese Bend, and (e) Palos Verdes. Inland lines in panels c and d show LiDAR analysis boundary, indicating possible additional cliff changes occurring further inland. See Fig. 4 for mapped locations.
Taller cliffs experienced more overall erosion (Fig. 6c), but cliff face retreat rates were generally uninfluenced by cliff height except for low height cliffs (0–10 m) that retrograded slower than all other cliffs (Fig. 6b). The reason for the small retreat rates of low cliffs is unknown, but could be from erosion-resistant low cliffs that sometimes lack the weaker upper Quaternary layer (for example in the Monterey area). At 5 m compartment scale, cliff height was not correlated with recent cliff retreat or erosion. However when retreat rates from unarmored cliffs were binned by height (10 m bins), the mean cliff height for each bin was significantly correlated with mean erosion rate when bins with fewer than 1000 observations were excluded ($r^2 = 0.71$, Fig. 6c). This correlation between mean binned cliff retreat rates and unarmored cliff height decreased when bins with fewer observations were considered.

5.2. Comparison to historical retreat rates

Distributions of historical and recent retreat rates were all skewed towards smaller magnitude events, but historical retreat rates were more evenly distributed and included higher percentages of large magnitude retreat rates (Fig. 7). The more evenly distributed historical cliff retreat could be because longer time periods average out the stochastic nature of cliff retreat, and suggests the system moves towards more spatially uniform retreat rates with time. 12.6, 4.7, and 1.3% of historical cliff top, recent cliff top, and recent cliff face retreat rates, respectively, exceeded −0.5 m/yr. Mean recent cliff top and cliff face retreat rates were lower than mean historical cliff top retreat rates by 52% and 83%, respectively (mean rates include zero change locations). The reason for the substantially lower rates is unknown but could be related to extensive anthropogenic alterations to the coastal system, different overall time scales (~70 vs. ~10 years), the episodic nature of cliff retreat, changes in ocean or subaerial forcing between these time periods, and/or variable quality of data sources. Additional retreat measurements over a wider variety of timescales are needed to test for timescale-dependence of erosion rates observed in other natural systems with intermittent erosion events (e.g. Finnegan et al., 2014; Ganti et al., 2016). However, in these other systems erosion rates are generally higher over shorter time intervals, opposite to the smaller decadal scale rates observed here. This opposition could be from the large spatial area evaluated here that can reduce the timescale-dependence of erosion rates (Sadler and Jerolmack, 2015; Ganti et al., 2016). Historical and recent maximum retreat rates were of similar magnitude.

Overall comparison of historical and recent individual compartment retreat rates (Fig. 8a, b) suggests they are not well correlated. However, historical and recent retreat rates are significantly ($p < 0.01$) inversely correlated for areas that experienced relatively large historical or recent cliff retreat, such that recent retreat rate decreases with elevated historical rate. The strength of the inverse correlation increases with cliff change magnitudes up to $r^2 = 0.91$ (Fig. 8d, e). This result suggests that using site-specific historical retreat rates to predict or project future decadal scale cliff retreat (and possibly longer time scales) using historical data could be problematic because of the varying timescales, forcing mechanisms, and system feedbacks. The stochastic nature of cliff retreat also complicates predictions when time scales vary and the time elapsed since previous failures is not considered. The inverse correlation is probably driven by geomorphic feedbacks, such that after a large cliff top failure, the cliff becomes relatively inactive from (1) a decrease in cliff slope and (2) increased wave protection from talus (and resulting beaches, when the cliffs contain sufficient beach-size sediment). These findings are consistent with Lee (2008) who found that extrapolating historical retreat is problematic without incorporating cliff-beach dynamics and forcing mechanisms over the observation period.
6. Summary

LiDAR data collected in 1998 and 2009–2010 were used to measure decadal scale coastal cliff erosion and cliff top retreat from the Mexico/California border to Bodega Head, California. Cliff face erosion was detected along 44% of the 595 km of cliffs evaluated. Mean cliff face and cliff top retreat rates were $-0.04$ and $-0.12$ m/yr, respectively, notably lower than historical (1929–1934 to 1998/2002) retreat rates of $-0.25$ m/yr. The lower recent rates could be from anthropogenic changes, varying time periods and forcing mechanisms, the stochastic nature of cliff retreat, and variable quality data sources. Distributions of historical retreat rates were spread more evenly and included higher percentages of large magnitude retreat rates, probably because the longer time span captured more locations that experienced episodic large cliff retreat events, thus spatially averaging out retreat rates. Large magnitude historical cliff retreat rates were inversely correlated with recent cliff retreat rates, suggesting possible problems with using historical retreat rates to project future cliff positions. The inverse correlation is probably driven by the episodic nature of large cliff retreat events, geomorphic feedbacks, and the cliff retreat cycle.

Maximum landward cliff top and cliff face retreat rates were about 4 m/yr, similar to maximum historical retreat rates. Localized high rates of coastal cliff change were found in Palos Verdes, Daly City, San Onofre State Beach, Point Reyes National Seashore, and Martin’s Beach and were often related to deep-seated and/or complex coastal landslides. Alongshore metrics of wave-cliff impact, precipitation, and cliff hardness generally did not correlate well with recent cliff retreat rates. On average, unarmored cliff faces retreated about three times further than armored cliffs. Although beaches can prevent wave-driven erosion, cliffs fronted by beaches retreated faster than those without beaches. However, the influence of beach width was not considered here, and narrow beaches that provide little erosion protection might have influenced this outcome. These findings highlight the need to monitor beach and cliff changes concurrently and examine the protective role of beaches in cliff processes.

The difference between cliff face change and cliff top retreat were used to quantify cliff steeping and establish a cliff top hazard index. Locations with relatively large cliff steeping and cliff top hazard values include San Onofre State Beach, Big Sur, Martin’s Beach, Daly City, and near Double Point. Additional surveys are needed to test the hazard index and identify hazard thresholds and probable timing of future failures.

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Fig. 7. Binned (0.1 m) distribution of historical and recent cliff retreat rates. Negative values indicate landward retreat.

Fig. 8. Comparison of (a) historical (1929–1934 to 1998/2002) cliff top and recent (1998/2002 to 2009–2010) cliff top retreat, (b) historical cliff top and recent cliff face retreat, and (c) recent cliff top and recent cliff face retreat. Color bars represent grid point density. (d–f) Correlations of retreat rates of corresponding top panels (a–c) for alongshore compartments with minimum historical or recent retreat rate magnitudes.


