

Article Title - La Palma Earthquakes

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8 **Abstract**

9 In September 2021, a significant jump in seismic activity on the island of La Palma (Ca-
 10 nary Islands, Spain) signaled the start of a volcanic crisis that still continues at the time
 11 of writing. Earthquake data is continually collected and published by the Instituto Ge-
 12 ográfico Nacional (IGN). We have created an accessible dataset from this and completed
 13 preliminary data analysis which shows seismicity originating at two distinct depths, con-
 14 sistent with the model of a two reservoir system feeding the currently very active vol-
 15 cano.

16 **1 Introduction**

17 The content of your notebook may be broken into any number of markdown or
 18 code cells. Markdown cells use MyST markdown and support standard markdown
 19 typography and many directives and roles for figures, tables, equations, etc.

20 La Palma is one of the west most islands in the Volcanic Archipelago of the Ca-
 21 nary Islands, a Spanish territory situated in the Atlantic Ocean where at their closest
 22 point are 100km from the African coast Figure 1 The island is one of the youngest, re-
 23 mains active and is still in the island forming stage.

24 Figures may be added to your notebook using the figure directive. They may re-
 25 fer to images saved in your `images/` folder, images from the web, or notebook cell
 26 outputs referenced by label. The `:name:` is used to reference the figure in your
 27 text; a reference to the following figure is found in the paragraph above. The fig-
 28 ure caption is given as the body of this directive.



Figure 1. Map of La Palma in the Canary Islands. Image credit NordNordWest

29 La Palma has been constructed by various phases of volcanism, the most recent and
 30 currently active being the *Cumbre Vieja* volcano, a north-south volcanic ridge that con-
 31 stitutes the southern half of the island.

32 **1.1 Eruption History**

33 A number of eruptions were recorded since the colonization of the islands by Eu-
 34 ropeans in the late 1400s, these are summarized in Table 1.

Table 1. Recent historic eruptions on La Palma

Name	Year
Current	2021
Teneguía	1971
Nambroque	1949
El Charco	1712
Volcán San Antonio	1677
Volcán San Martín	1646
Tajuya near El Paso	1585
Montaña Quemada	1492

35 Simple tables may be created using the list-table directive. Similar to figures, ta-
 36 bles may be referenced in the text by their **name**. The caption for this table is the
 37 first line of the directive.

38 This equates to an eruption on average every 79 years up until the 1971 event. The
 39 probability of a future eruption can be modeled by a Poisson distribution (1).

40 Numbered equations may be defined using the math directive or in line. Equa-
 41 tions defined with the math directive may be reference in the text by label.

$$p(x) = \frac{e^{-\lambda} \lambda^x}{x!} \quad (1)$$

42 Where λ is the number of eruptions per year, $\lambda = \frac{1}{79}$ in this case. The probabil-
 43 ity of a future eruption in the next t years can be calculated by:

$$p_e = 1 - e^{-t\lambda} \quad (2)$$

44 So following the 1971 eruption the probability of an eruption in the following 50
 45 years — the period ending this year — was 0.469. After the event, the number of erup-
 46 tions per year moves to $\lambda = \frac{1}{75}$ and the probability of a further eruption within the next
 47 50 years (2022-2071) rises to 0.487 and in the next 100 years, this rises again to 0.736.

48 1.2 Magma Reservoirs

49 You may add citations two ways. First, you may simply insert a markdown link
 50 to a DOI like so: Thompson et al. (1994). No additional bibliographic informa-
 51 tion is required for this approach; the reference will be looked up by DOI and added
 52 implicitly to the references. Alternatively, you may provide the bibliography di-
 53 rectly as `references.bib` BibTeX file, then embed the citation by BibTeX key
 54 in your text using the `@cite2023` or `[@cite2023; @cite2023b]` for narrative or
 55 parenthetical citations, respectively. The following paragraph provides an exam-
 56 ple of this. A single paper may combine both DOI and BibTeX citations.

57 Studies of the magma systems feeding the volcano, such as Marrero et al. (2019)
 58 has proposed that there are two main magma reservoirs feeding the Cumbre Vieja vol-
 59 cano; one in the mantle (30-40km depth) which charges and in turn feeds a shallower
 60 crustal reservoir (10-20km depth).

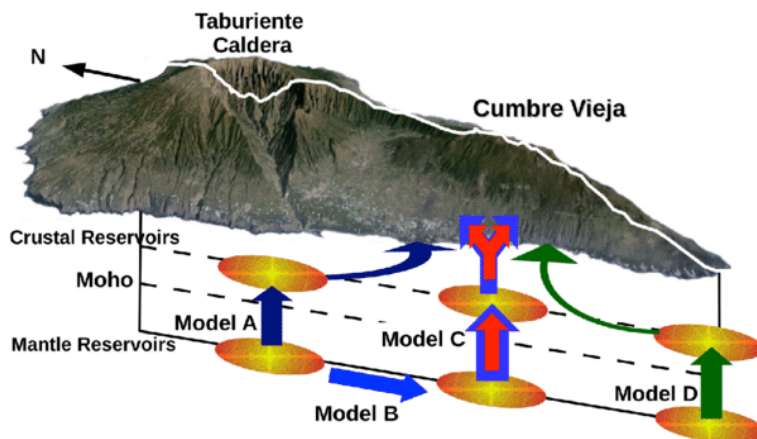


Figure 2. Proposed model from Marrero et al

61 In this paper, we look at recent seismicity data to see if we can see evidence of such
 62 a system action, see Figure 2.

63 2 Dataset

64 All data used in the notebook should be present in the `data/` folder so notebooks
 65 may be executed in place with no additional input.

66 The earthquake dataset used in our analysis was generated from the IGN web portal
 67 this is public data released under a permissive license. Data recorded using the network
 68 of Seismic Monitoring Stations on the island. A web scraping script was developed
 69 to pull data into a machine-readable form for analysis. That code tool is available on GitHub
 70 along with a copy of recently updated data.

71 2.1 Main Timeline Figure

72 Code cells may be seamlessly interleaved with markdown cells. Currently, with
 73 a single-article submission, code cannot be hidden in the output document.

```
74 import pandas as pd
75 import matplotlib
76 import matplotlib.pyplot as plt
77 %matplotlib inline
78 import seaborn as sns
79 import numpy as np
80 sns.set_theme(style="whitegrid")
```

```
81 def make_category_columns(df):
82     df['Depth'] = 'Shallow (<18km)'
83     df.loc[(df['Depth(km)'] >= 18) & (df['Depth(km)'] <= 28), 'Depth'] = 'Interchange (18km>x>28)'
84     df.loc[df['Depth(km)'] >= 28, 'Depth'] = 'Deep (>28km)'
85
86     df['Mag'] = 0
87     df.loc[(df['Magnitude'] >= 1) & (df['Magnitude'] <= 2), 'Mag'] = 1
88     df.loc[(df['Magnitude'] >= 2) & (df['Magnitude'] <= 3), 'Mag'] = 2
```

```

89     df.loc[(df['Magnitude'] >= 3) & (df['Magnitude'] <= 4), 'Mag'] = 3
90     df.loc[(df['Magnitude'] >= 4) & (df['Magnitude'] <= 5), 'Mag'] = 4
91
92     return df

```

93 2.2 Visualising Long term earthquake data

94 Data taken directly from the IGN Catalog

95 Supported cell outputs below include pandas dataframe, raw text output, matplotlib
96 plot, and seaborn plot.

```

97 df_ign = pd.read_csv('./data/lapalma_ign.csv')
98 df_ign = make_category_columns(df_ign)
99 df_ign.head()

```

Event	Date	Time	Latitude	Longitude	Depth(km)	\
0	es2017eugju	2017 -03 -09 23:44:06	28.5346	-17.8349	26.0	
1	es2017euhlh	2017 -03 -10 00:16:10	28.5491	-17.8459	27.0	
2	es2017cpaoh	2017 -03 -10 00:16:11	28.5008	-17.8863	20.0	
3	es2017eunnk	2017 -03 -10 03:20:26	28.5204	-17.8657	30.0	
4	es2017kajei	2017 -08 -21 02:06:55	28.5985	-17.7156	0.0	

	Intensity	Magnitude	Type	Mag	Location	\
0		1.6	4	NE FUENCALIENTE DE LA PALMA.IL		
1		2.0	4	N FUENCALIENTE DE LA PALMA.ILP		
2		2.1	4	W LOS CANARIOS.ILP		
3		1.6	4	NW FUENCALIENTE DE LA PALMA.IL		
4		1.6	4	E EL PUEBLO.ILP		

	DateTime	Timestamp	Swarm	Phase	\
0	2017 -03 -09 23:44:06	1489103046000000000	0.0	0	
1	2017 -03 -10 00:16:10	1489104970000000000	0.0	0	
2	2017 -03 -10 00:16:11	1489104971000000000	0.0	0	
3	2017 -03 -10 03:20:26	1489116026000000000	0.0	0	
4	2017 -08 -21 02:06:55	1503281215000000000	0.0	0	

	Depth	Mag
0	Interchange (18km>x>28km)	1
1	Interchange (18km>x>28km)	2
2	Interchange (18km>x>28km)	2
3	Deep (>28km)	1
4	Shallow (<18km)	1

```

127 df_ign['DateTime'] = pd.to_datetime(df_ign['Date'] + ' ' + df_ign['Time'])
128 df_ign['DateTime'];

```

```

129 df_ign_early = df_ign[df_ign['DateTime'] < '2021 -09 -11']
130 df_ign_pre = df_ign[(df_ign['DateTime'] >= '2021 -09 -11') & (df_ign['DateTime'] < '2021 -09 -19')]
131 df_ign_phase1 = df_ign[(df_ign['DateTime'] >= '2021 -09 -19 14:13:00') & (df_ign['DateTime'] < '2021 -10 -01')]
132 df_ign_phase2 = df_ign[(df_ign['DateTime'] >= '2021 -10 -01') & (df_ign['DateTime'] < '2021 -12 -01')]
133 df_ign_phase3 = df_ign[(df_ign['DateTime'] >= '2021 -12 -01') & (df_ign['DateTime'] <= '2021 -12 -31')]
134
135 df_erupt = df_ign[(df_ign['Date'] < '2022 -01 -01') & (df_ign['Date'] > '2021 -09 -11')]
136

```

```

137 df_erupt_1 = df_erupt[df_erupt['Magnitude'] < 1.0]
138 df_erupt_2 = df_erupt[(df_erupt['Magnitude'] >= 1.0)&(df_erupt['Magnitude'] < 2.0)]
139 df_erupt_3 = df_erupt[(df_erupt['Magnitude'] >= 2.0)&(df_erupt['Magnitude'] < 3.0)]
140 df_erupt_4 = df_erupt[(df_erupt['Magnitude'] >= 3.0)&(df_erupt['Magnitude'] < 4.0)]
141 df_erupt_5 = df_erupt[df_erupt['Magnitude'] > 4.0]

142 tab20_colors = (
143     (0.12156862745098039, 0.4666666666666667, 0.7058823529411765 ), # 1f77b4
144     (0.6823529411764706, 0.7803921568627451, 0.9098039215686274 ), # aec7e8
145     (1.0, 0.4980392156862745, 0.054901960784313725), # ff7f0e
146     (1.0, 0.7333333333333333, 0.47058823529411764 ), # ffbb78
147     (0.17254901960784313, 0.6274509803921569, 0.17254901960784313 ), # 2ca02c
148     (0.596078431372549, 0.8745098039215686, 0.5411764705882353 ), # 98df8a
149     (0.8392156862745098, 0.15294117647058825, 0.1568627450980392 ), # d62728
150     (1.0, 0.596078431372549, 0.5882352941176471 ), # ff9896
151     (0.5803921568627451, 0.403921568627451, 0.7411764705882353 ), # 9467bd
152     (0.7725490196078432, 0.6901960784313725, 0.8352941176470589 ), # c5b0d5
153     (0.5490196078431373, 0.33725490196078434, 0.29411764705882354 ), # 8c564b
154     (0.7686274509803922, 0.611764705882353, 0.5803921568627451 ), # c49c94
155     (0.8901960784313725, 0.4666666666666667, 0.7607843137254902 ), # e377c2
156     (0.9686274509803922, 0.7137254901960784, 0.8235294117647058 ), # f7b6d2
157     (0.4980392156862745, 0.4980392156862745, 0.4980392156862745 ), # 7f7f7f
158     (0.7803921568627451, 0.7803921568627451, 0.7803921568627451 ), # c7c7c7
159     (0.7372549019607844, 0.7411764705882353, 0.13333333333333333 ), # bcbd22
160     (0.8588235294117647, 0.8588235294117647, 0.5529411764705883 ), # dbdb8d
161     (0.09019607843137255, 0.7450980392156863, 0.8117647058823529 ), # 17becf
162     (0.6196078431372549, 0.8549019607843137, 0.8980392156862745), # 9edae5
163 )

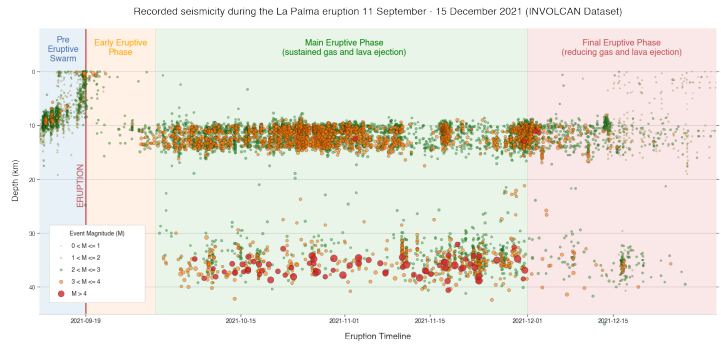
164 from matplotlib.patches import Rectangle
165
166 import datetime as dt
167 from matplotlib.dates import date2num, num2date
168
169 matplotlib.rcParams['font.family'] = "sans -serif"
170 matplotlib.rcParams['xtick.labelsize'] = 14
171 matplotlib.rcParams['ytick.labelsize'] = 14
172 matplotlib.rcParams['ytick.labelleft'] = True
173 matplotlib.rcParams['ytick.labelright'] = True
174
175 %matplotlib inline
176 fig = matplotlib.pyplot.figure(figsize=(24,12))
177 fig.tight_layout()
178 # Creating axis
179 # add_axes([xmin,ymin,dx,dy])
180 ax_min = fig.add_axes([0.01, 0.01, 0.01, 0.01])
181 ax_min.axis('off')
182 ax_max = fig.add_axes([0.99, 0.99, 0.01, 0.01])
183 ax_max.axis('off')
184
185 ax_timeline = fig.add_axes([0.04, 0.1, 0.92, 0.85])
186 ax_timeline.spines["top"].set_visible(False)
187 ax_timeline.spines["right"].set_visible(False)
188 ax_timeline.spines["left"].set_visible(False)
189 ax_timeline.grid(axis='x')

```

```

190
191
192 ax_timeline.axvline(x=dt.datetime(2021, 9, 19, 14, 13), ymin=0.075, ymax=0.98, color='r', linewidth=2)
193
194
195 def make_scatter(df, c, alpha=0.8):
196     M = 3*np.exp2(1.3*df['Magnitude'])
197     return ax_timeline.scatter(df['DateTime'], df['Depth(km)'], s=M, c=c, alpha=alpha, edgecolor='black')
198
199 # make_scatter(df_erupt, c=tab20c_colors[ -1])
200 points_1 = make_scatter(df_erupt_1, c=[tab20_colors[12]], alpha=0.3)
201 points_2 = make_scatter(df_erupt_2, c=[tab20_colors[16]], alpha=0.4)
202 points_3 = make_scatter(df_erupt_3, c=[tab20_colors[4]], alpha=0.5)
203 points_4 = make_scatter(df_erupt_4, c=[tab20_colors[2]], alpha=0.6)
204 points_5 = make_scatter(df_erupt_5, c=[tab20_colors[6]], alpha=0.8)
205
206 ax_timeline.tick_params(axis='x', labelrotation=0, bottom=True)
207 ax_timeline.set_ylabel('')
208 ax_timeline.yaxis.set_ticks_position('both')
209 ax_timeline.yaxis.set_ticks_position('both')
210
211 xticks = ax_timeline.get_xticks()
212 new_xticks = [date2num(pd.to_datetime('2021 -09 -11')),
213              date2num(pd.to_datetime('2021 -09 -19 14:13:00'))]
214 new_xticks = np.append(new_xticks, xticks[2: -1])
215 ax_timeline.set_xticks(new_xticks)
216
217 ax_timeline.invert_yaxis()
218 ax_timeline.spines['bottom'].set_position(('data', 45))
219 ax_timeline.margins(tight=True, x=0)
220 ax_timeline.legend(
221     [points_1, points_2, points_3, points_4, points_5],
222     ['0 < M <= 1', '1 < M <= 2', '2 < M <= 3', '3 < M <= 4', 'M > 4'],
223     loc='lower left', bbox_to_anchor=(0.01, 0.1, 0.15, 0.1), fancybox=True,
224     borderpad=1.0, labelspace=1, mode="expand", title="Event Magnitude (M)",
225     fontsize=14, title_fontsize=14, framealpha=1)
226
227 ax_timeline.set_ylim(ax_timeline.get_ylim()[0], -9)
228
229 plt.annotate('ERUPTION', (0.055, 0.42), rotation=90, xycoords='axes fraction', fontweight='bold')
230 plt.annotate('Pre\nEruptive\nSwarm', (0.035, 0.88), rotation=0, xycoords='axes fraction', fontweight='bold')
231 plt.annotate('Early Eruptive\nPhase', (0.12, 0.9), rotation=0, xycoords='axes fraction', fontweight='bold')
232 plt.annotate('Main Eruptive Phase\n(sustained gas and lava ejection)', (0.45, 0.9), rotation=0, fontweight='bold')
233 plt.annotate('Final Eruptive Phase\n(reducing gas and lava ejection)', (0.86, 0.9), rotation=0, fontweight='bold')
234
235 ax_timeline.add_patch(Rectangle((date2num(pd.to_datetime('2021 -09 -11')), -8), date2num(pd.to_datetime('2021 -09 -19 14:13:00')), -8), date2num(pd.to_datetime('2021 -09 -11')), -8))
236 ax_timeline.add_patch(Rectangle((date2num(pd.to_datetime('2021 -09 -19 14:13:00')), -8), date2num(pd.to_datetime('2021 -10 -01')), -8), date2num(pd.to_datetime('2021 -09 -19 14:13:00')), -8))
237 ax_timeline.add_patch(Rectangle((date2num(pd.to_datetime('2021 -10 -01')), -8), date2num(pd.to_datetime('2021 -12 -01')), -8), date2num(pd.to_datetime('2021 -10 -01')), -8))
238 ax_timeline.add_patch(Rectangle((date2num(pd.to_datetime('2021 -12 -01')), -8), date2num(pd.to_datetime('2021 -12 -01')), -8), date2num(pd.to_datetime('2021 -12 -01')), -8))
239
240 ax_timeline.set_title("Recorded seismicity during the La Palma eruption 11 September - 15 December 2021")
241 ax_timeline.set_ylabel("Depth (km)", dict(fontsize=20), labelpad=20)
242 ax_timeline.set_xlabel("Eruption Timeline", dict(fontsize=20), labelpad=20);

```



243

244

2.3 Cumulative Distrubtion Plots

245

```
def cumulative_events_mag_depth(df, hue='Depth', kind='scatter', ax=None, dpi=100, palette=None
```

246

```
matplotlib.rcParams['ytick.labelright'] = False
```

247

```
g = sns.jointplot(x="Magnitude", y="Depth(km)", data=df,
```

248

```
kind=kind, hue=hue, height=10, space=0.1, marginal_ticks=False, ratio=8, a
```

249

```
hue_order=['Shallow (<18km)', 'Interchange (18km>x>28km)', 'Deep (>28km)']
```

250

```
ax=ax, palette=palette, ylim=(-2,50), xlim=(0.3,5.6), edgecolor=".2", ma
```

251

```
if kde:
```

252

```
g.plot_joint(sns.kdeplot, color="b", zorder=1, levels=15, ax=ax)
```

253

```
g.fig.axes[0].invert_yaxis();
```

254

```
g.fig.set_dpi(dpi)
```

255

```
import warnings
```

256

257

```
with warnings.catch_warnings():
```

258

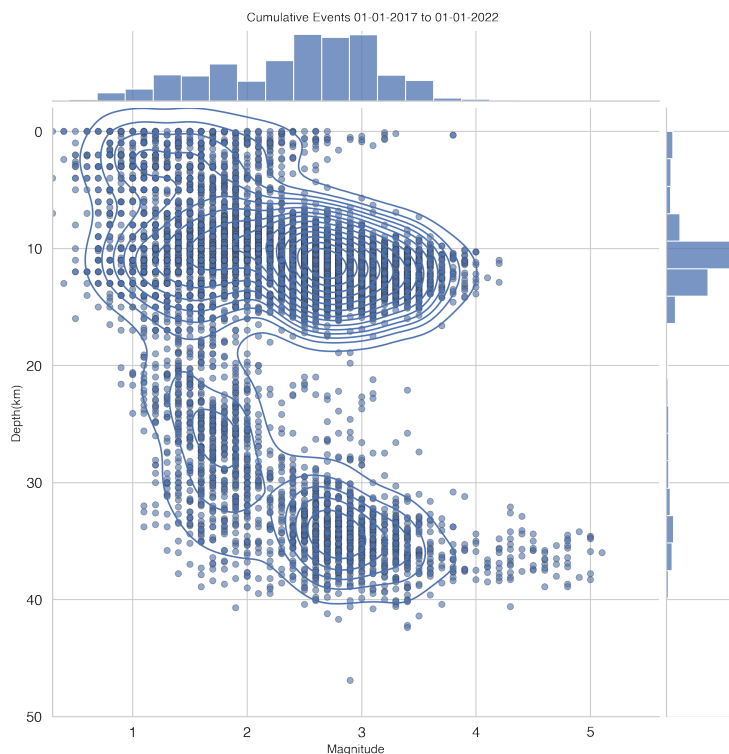
```
warnings.simplefilter("ignore")
```

259

```
cumulative_events_mag_depth(df_ign, hue=None, dpi=200)
```

260

```
plt.suptitle('Cumulative Events 01 -2017 to 01 -01 -2022', y=1.01);
```

261

3 Results

262

263 The dataset was loaded into this Jupyter notebook and filtered down to La Palma
 264 events only. This results in 5465 data points which we then visualized to understand their
 265 distributions spatially, by depth, by magnitude and in time.

266 From our analysis above, we can see 3 different systems in play.

267 Firstly, the shallow earthquake swarm leading up to the eruption on 19th September,
 268 related to significant surface deformation and shallow magma intrusion.

269 After the eruption, continuous shallow seismicity started at 10-15km correspond-
 270 ing to magma movement in the crustal reservoir.

271 Subsequently, high magnitude events begin occurring at 30-40km depths correspond-
 272 ing to changes in the mantle reservoir. These are also continuous but occur with a lower
 273 frequency than in the crustal reservoir.

4 Conclusions

274

275 From the analysis of the earthquake data collected and published by IGN for the
 276 period of 11 September through to 9 November 2021. Visualization of the earthquake
 277 events at different depths appears to confirm the presence of both mantle and crustal
 278 reservoirs as proposed by Marrero et al. (2019).

Open Research

279

280 A web scraping script was developed to pull data into a machine-readable form for
 281 analysis. That code tool is available on GitHub along with a copy of recently updated
 282 data.

283 **References**

- 284 Marrero, J., García, A., Berrocoso, M., Llinares, Á., Rodríguez-Losada, A., & Ortiz,
285 R. (2019, 7). Strategies for the development of volcanic hazard maps in mono-
286 genetic volcanic fields: the example of La Palma (Canary Islands). *Journal of*
287 *Applied Volcanology*, 8. doi: 10.1186/s13617-019-0085-5
- 288 Thompson, J. D., Higgins, D. G., & Gibson, T. J. (1994). Clustal w: improv-
289 ing the sensitivity of progressive multiple sequence alignment through sequence
290 weighting, position-specific gap penalties and weight matrix choice. *Nucleic Acids*
291 *Research*, 22(22), 4673–4680. Retrieved from [http://dx.doi.org/10.1093/nar/](http://dx.doi.org/10.1093/nar/22.22.4673)
292 [22.22.4673](http://dx.doi.org/10.1093/nar/22.22.4673) doi: 10.1093/nar/22.22.4673