



Multi-model probability assessments in the Long-Valley volcanic region (CA)

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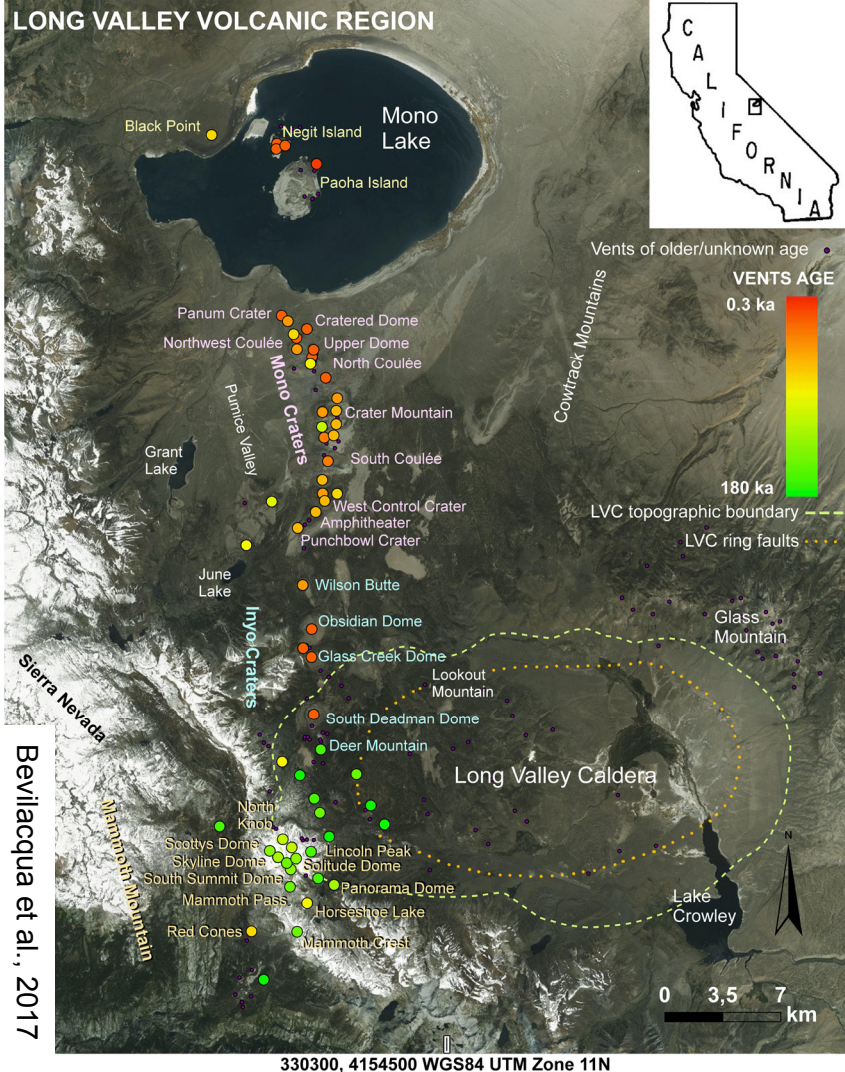


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LONG VALLEY VOLCANIC REGION



- The Long Valley volcanic region -

Long Valley caldera (LVC), was created by the eruption of $>650\text{km}^3$ tephra $\sim 760\text{ka}$ (Bishop tuff).

After 180ka, the eruptions have been mostly localized at **Mammoth Mountain**, and its periphery and along the **Mono-Inyo Craters** volcanic chain, stretching $\sim 45\text{km}$ North outside the caldera, towards Mono Lake.

The most recent period of **unrest started in 1978** - several seismic swarms in LVC and below Mammoth Mountain, and diffuse volcanic CO_2 emissions.

We produce **long-term forecasting models** for the timing and location of future eruptions, with uncertainty.

The models are **doubly stochastic**, i.e. each sample is made in two steps:

- A) the random choice of the epistemic uncertainty,
- B) the random determination of the forecasts, conditional on A).

All probability values will have **confidence intervals due to the uncertainties**.

- Eruptive record uncertainty model -

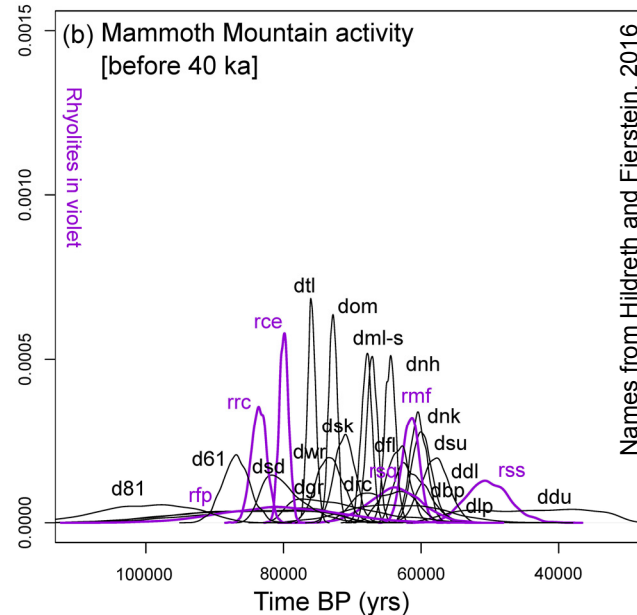
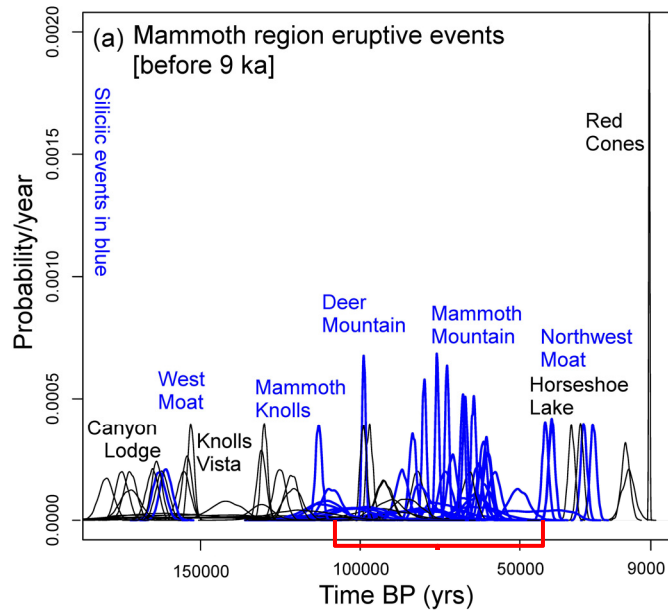
The radio-isotopic age of past eruptions is affected by uncertainty, but **constrained** by the stratigraphic sequence.

A Monte Carlo simulation samples the ages and **resamples those** which contradict stratigraphic constraints.

This enables the reconstruction of a **stochastic record** of 134 events after 180 ka including inter-event dependence.

Our definition of event is an eruptive activity which is **interrupted** by evidence of quiescence of yearly scale.

Sometimes multiple units can correspond to a single event, or single units can correspond to multiple events.



The first and older part of eruptive record is related to:
Mammoth region (81 events)

Mammoth Mountain
 26^{+3} events [110 - 40] ka

Mammoth periphery
52 events [180 - 9] ka

The figure is based on 10^4 samples.

- Eruptive record uncertainty model -

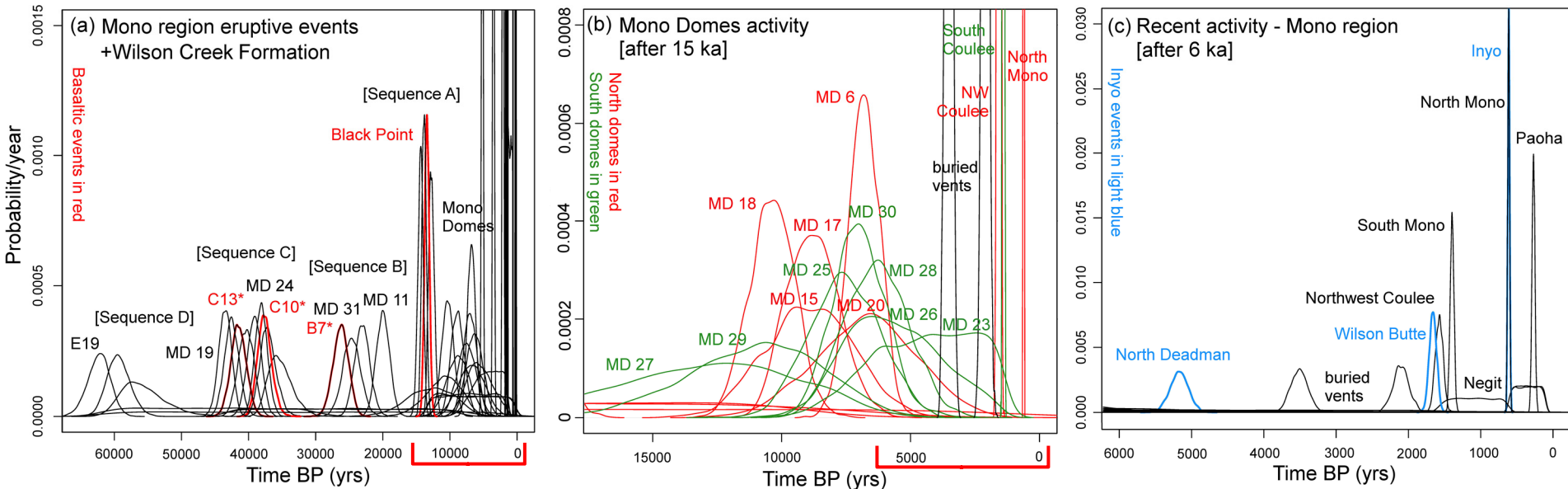
The second part of eruptive record is related to **Mono region (53 events)**

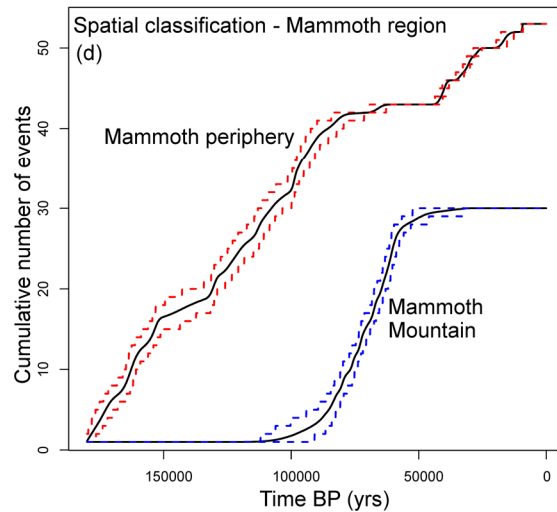
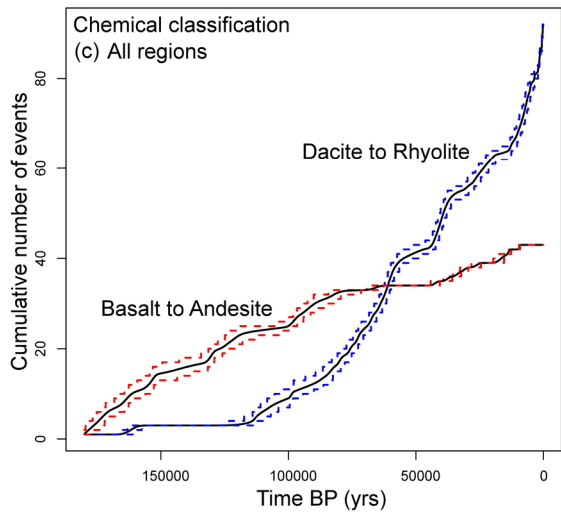
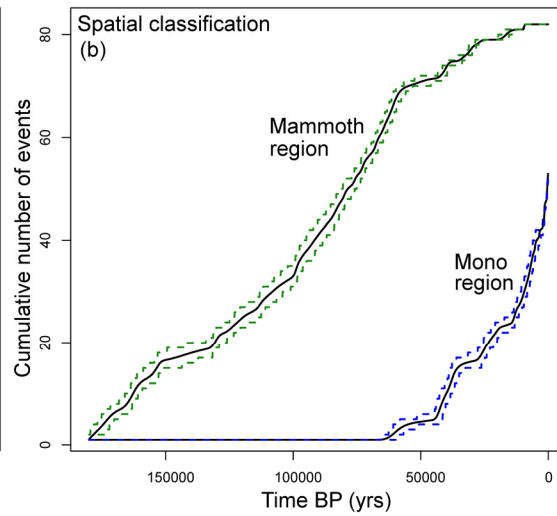
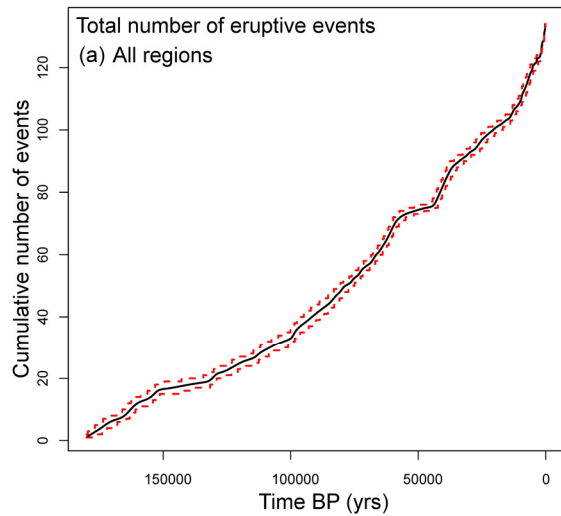
<i>Mono Lake</i>	8 events	[14 - 0.2] ka
<i>Inyo craters</i>	3 events	[5.5 - 0.6] ka
<i>Mono domes</i>	23 ⁺² events	[65 - 0.6] ka
<i>Wilson Creek Formation</i>	21 ⁻⁴ events	[65 - 13] ka

The **Wilson Creek Formation** preserves 5 sequences of ash beds, most of them tracing back to Mono-Inyo craters.

This enabled us to include **older eruptions** whose vents can be buried.

The figure is based on 10⁴ samples.





- Cumulative event number -

Mean curves, with 5%ile and 95%ile values, as a function of time. Based on 2500 samples.

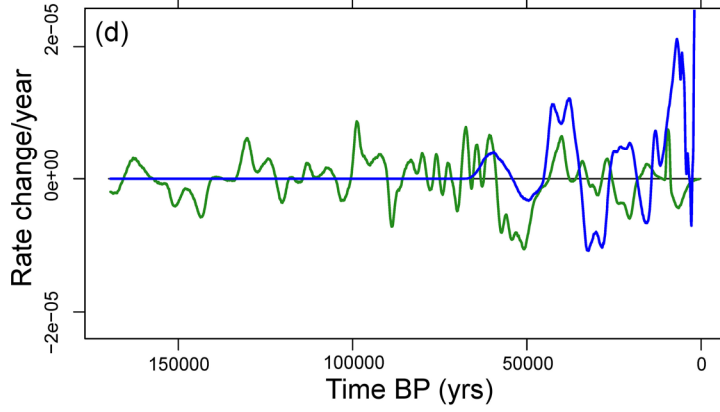
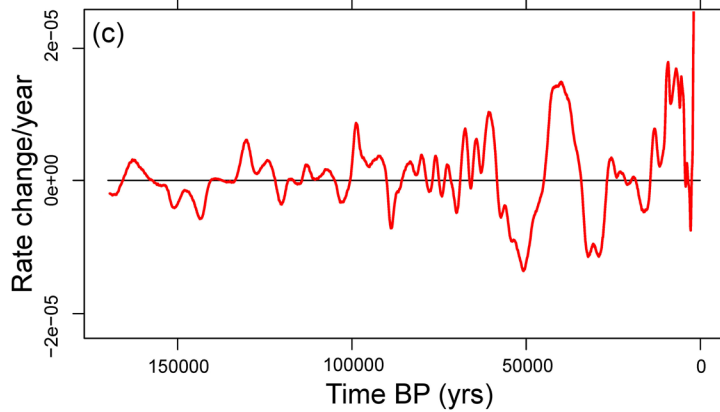
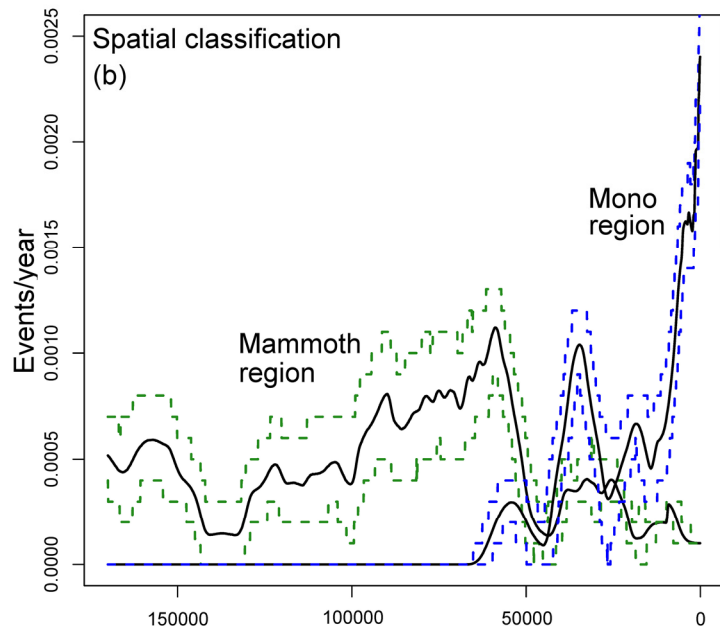
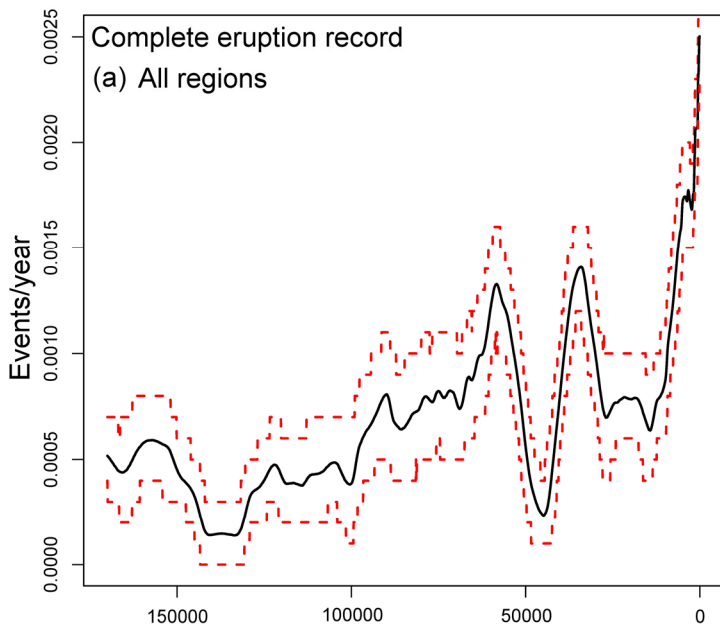
The slope of the plot (a) is **significantly regular**, except for a weak apparent acceleration of the activity.

Plot (b) shows the **increase of activity in Mono region** after 60 ka, complementary to an apparent decrease in Mammoth region.

In plot (c) the rate for the silicic events is approximately **twice the rate** of basaltic events.

In plot (d) Mammoth Mountain activity is mostly constrained between 100 ka and 50 ka.

At [65 - 40] ka can be seen **a stop** in the activity of the Mammoth periphery.



- Event rate -

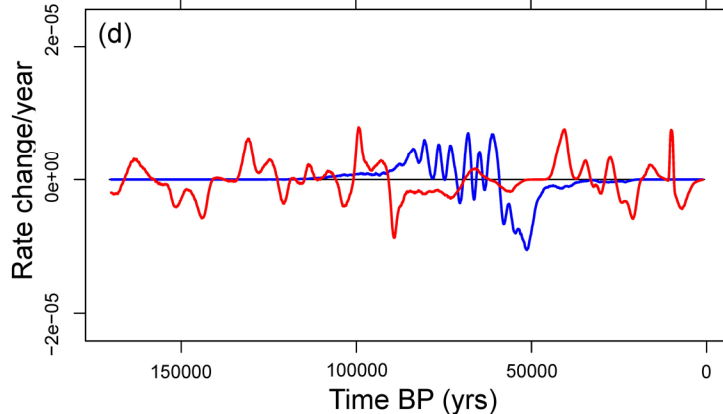
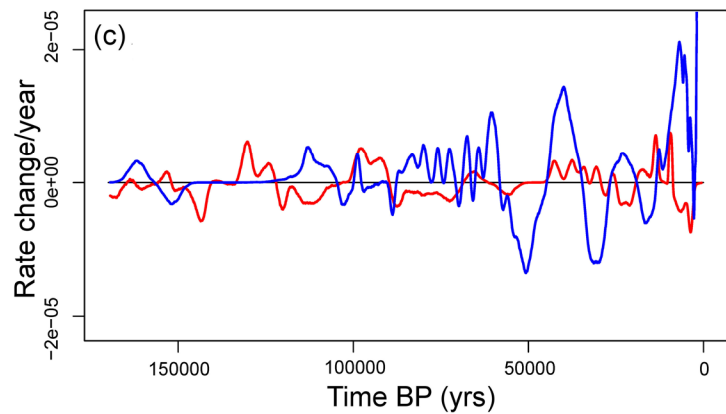
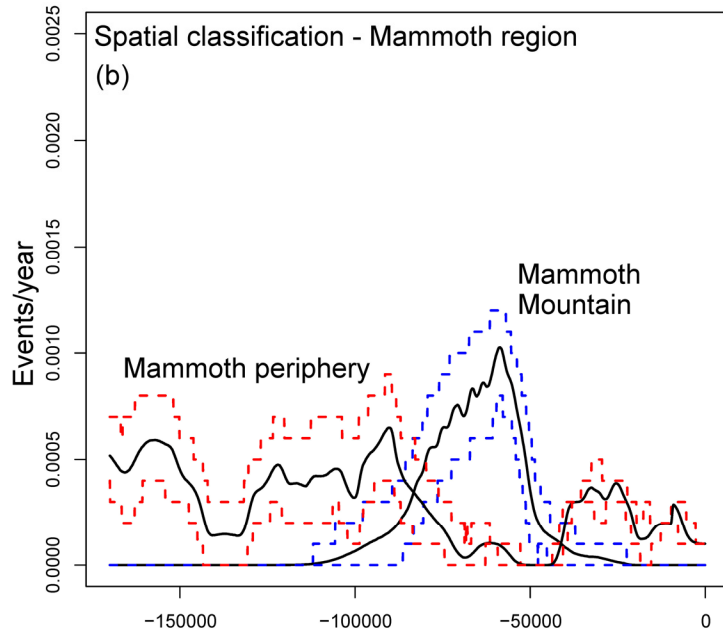
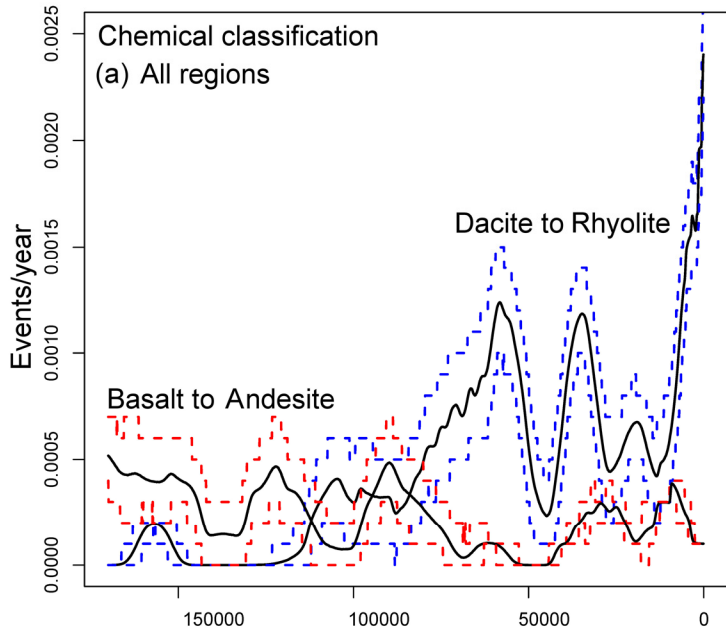
Including uncertainty.

In plot (a) after a rate reduction at 140-130 ka, there was **regular rise until 60 ka**, then a drop down and a new **burst at 40-30 ka**, reaching $1/750 \text{ yr}^{-1}$ rate.

After 15 ka, there was a burst **twice** the previous, to $1/400 \text{ yr}^{-1}$.

Plot (b) shows that the rise of new activity after 50 ka is mostly related to Mono region, in **four clusters**.

Rate change plots (c and d) show an **alternation** of negatively and positively valued intervals, **pulsing** with a $\sim 15\text{-}20 \text{ kyr}$ period.



- Event rate -

Including uncertainty.

In plot (a) **basaltic activity** has **4-5 clusters**, with a gap when the **silicic activity** rises to a maximum at 60 ka.

Three bursts in the more recent times are evident.

In plot (b) **decrease** of peripheral activity is clear at 80-40 ka.

In plot (c) an **alternation of basaltic and silicic activity** has peaks of in silicic activity corresponding to valleys in basaltic activity and vice versa.

Exceptions are before 130 ka and at 60-40 ka, with a more coupled behavior.

- Cox processes based on moving averages -

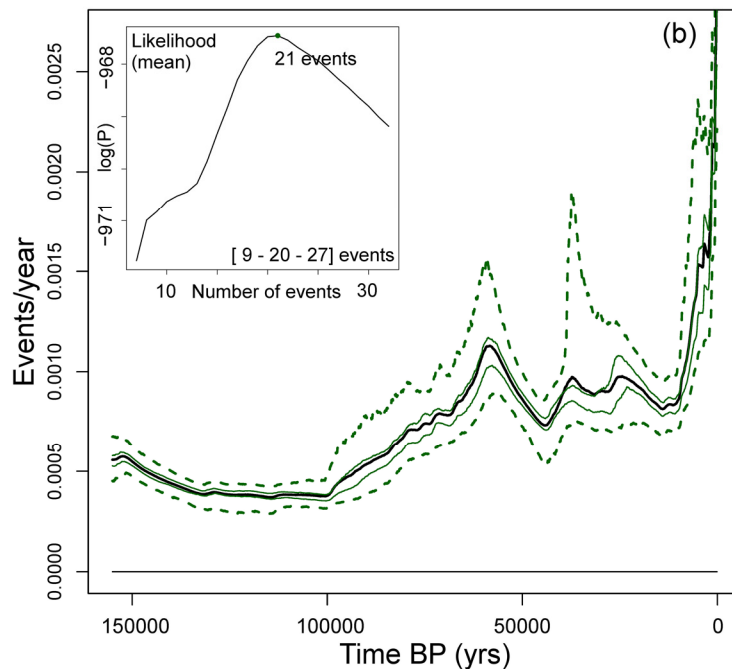
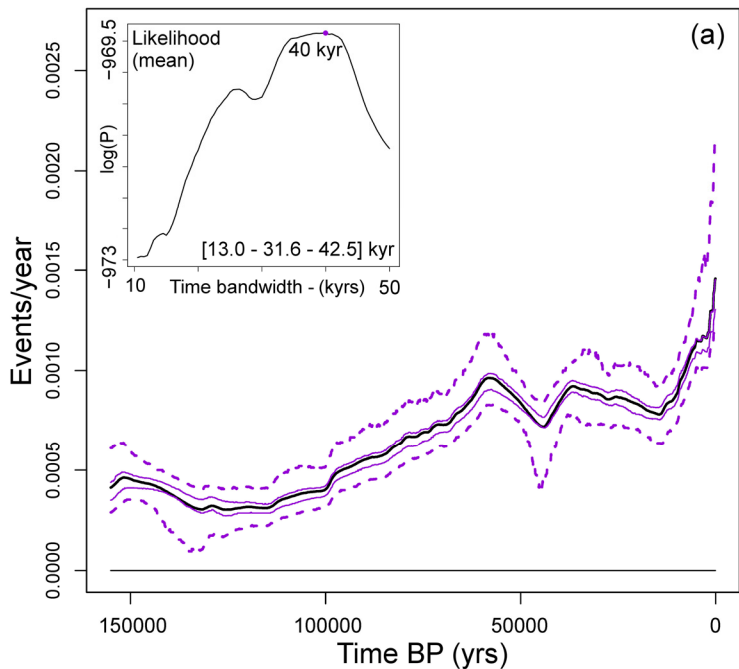
MODEL 1

The intensity function $\lambda(t)$, which is the average rate of our forecasting model at time t , is defined by the ratio $\#events / T$ on a left window $[t-T, t]$. T is the only parameter.

MODEL 1B - (adaptive time-window correction)

The intensity function $\lambda(t)$ is the ratio $k / (t - t_n - k)$, where k is a parameter, and t_n is the time of the last event occurred before time t .

This gives the potential for **higher spikes** of intensity. In particular, the current intensity values are two times as high as those in M1.



Mean curves, 5%ile, 33%ile, 67%ile, 95%ile values are displayed. Based on 1000 samples.

The bandwidth T and the number k are selected to **maximize likelihood (MLE)** of past events.

MLE can be done in the average (small boxes), or **sample-wise** inside a Monte Carlo simulation.

- Cox processes with self-excitement (Cox-Hawkes) -

The intensity function $\lambda(t)$ is the sum $\lambda_0 + f(t)$, where λ_0 is called base rate, and $f > 0$ is called self-excitement function.

Pre-existing events increase f with a jump Δf , and the effect of each contribution then decreases exponentially with time.

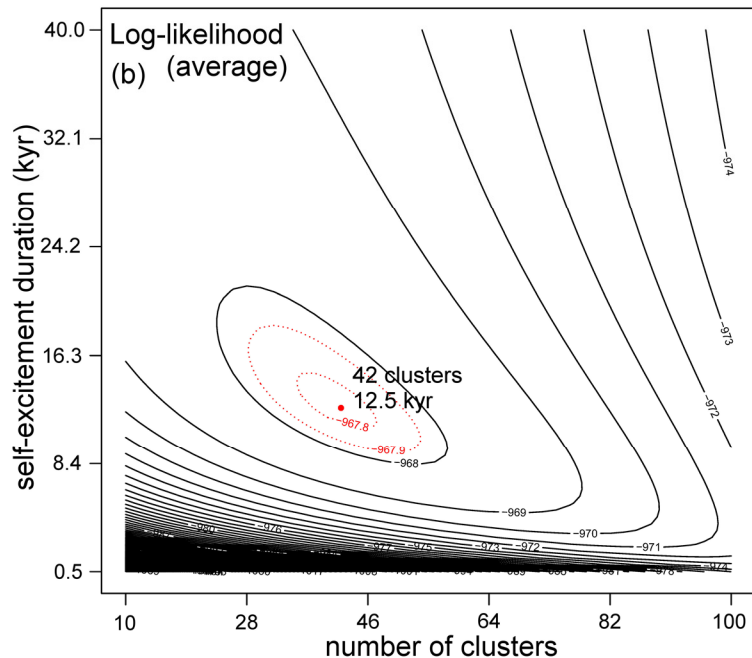
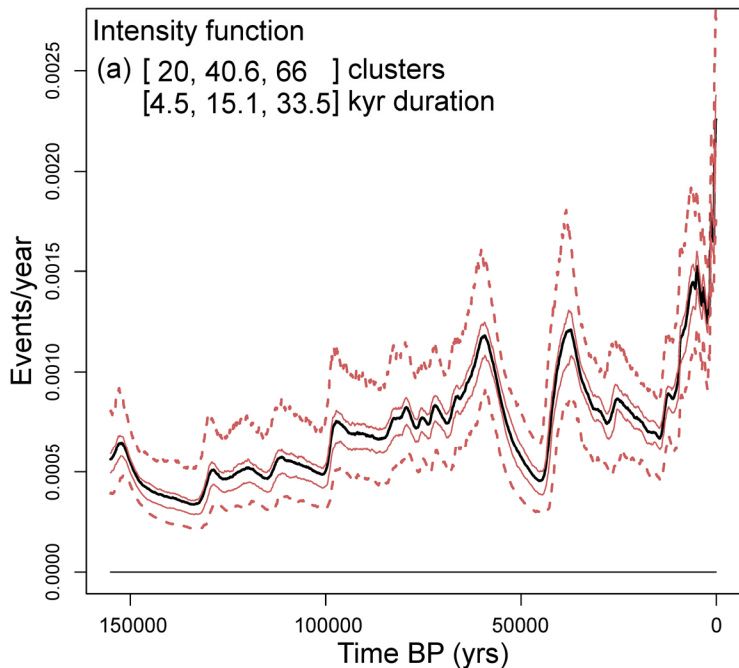
The parameter space is two dimensional:

number of clusters n ,

i.e. the number of ancestor events generated by the base rate λ_0

self-excitement duration τ ,

i.e. the time after an event before its intensity contribution Δf becomes negligible.



Higher spikes than in M1, but more frequent and shorter than in M1B.

A similar approach produced promising results on **Campi Flegrei** (Bevilacqua et al. 2016) and **Auckland volcanic field** (Bebbington and Cronin, 2011)

- Forecasting models results -

Our three models give **different forecasts**, with the current intensity $\lambda(t=0)$ smallest in M1 and biggest in M1B.

The table shows the probability estimates of an eruption in the **next 10 and 50 years**, with uncertainty.

These results are preliminary.

Two alternative **multi-model mixtures** are displayed, based on a maximum likelihood estimator (MLE), or a Bayesian Model Averaging (BMA), with consistent results.

(a)	Model 1			Model 1B			Model 2		
	5 th %ile	mean	95 th %ile	5 th %ile	mean	95 th %ile	5 th %ile	mean	95 th %ile
λ [yrs ⁻¹]	1/460	1/686	1/810	1/189	1/344	1/452	1/363	1/443	1/571
$P(\Delta t_{\text{next}} < 10 \text{ yrs})$	1.23%	1.45%	2.15%	2.19%	2.86%	5.16%	1.74%	2.23%	2.72%
$P(\Delta t_{\text{next}} < 50 \text{ yrs})$	5.99%	7.03%	10.31%	10.48%	13.51%	23.29%	8.38%	10.67%	12.86%
$P(\text{MLE} = M_x)$	0.4%			49.4%			50.2%		
BMA - score	0.6%	7.5%	22.1%	3.5%	43.8%	84.5%	9.6%	48.8%	95.7%

Based on 1000 samples.

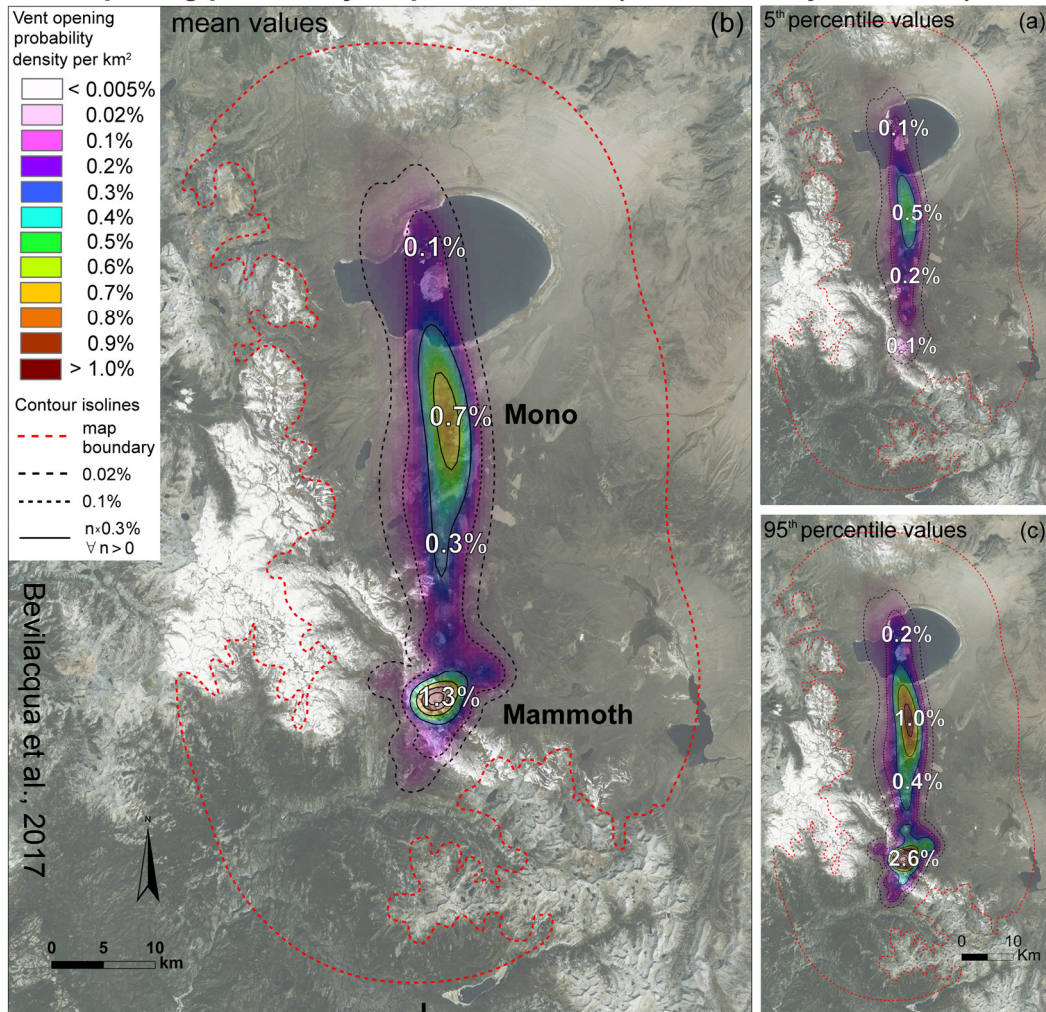
Both methods **penalize Model 1** while the others show similar performances.

(b)	MLE - selection			BMA		
	5 th %ile	mean	95 th %ile	5 th %ile	mean	95 th %ile
λ [yrs ⁻¹]	1/279	1/376	1/473	1/289	1/393	1/488
$P(\Delta t_{\text{next}} < 10 \text{ yrs})$	2.09%	2.62%	3.52%	2.03%	2.51%	3.39%
$P(\Delta t_{\text{next}} < 50 \text{ yrs})$	10.03%	12.45%	16.40%	9.73%	11.96%	15.86%

MLE approach always follows only the model which **maximizes performance**, sample wise.

BMA **linearly averages the models**, weighting in proportion to their hind-casting performance.

Vent opening probability maps - Model 1 (kernel density estimator)



- Vent opening maps -

A "map of vent opening" is the **spatial estimate** of the probability of vent opening per km² in each point.

This probability is **conditional** on the occurrence of a new eruption, without a temporal window.

The forecasts are affected by uncertainty: we calculate the **mean, 5th and 95th percentile** values of the vent opening pdf.

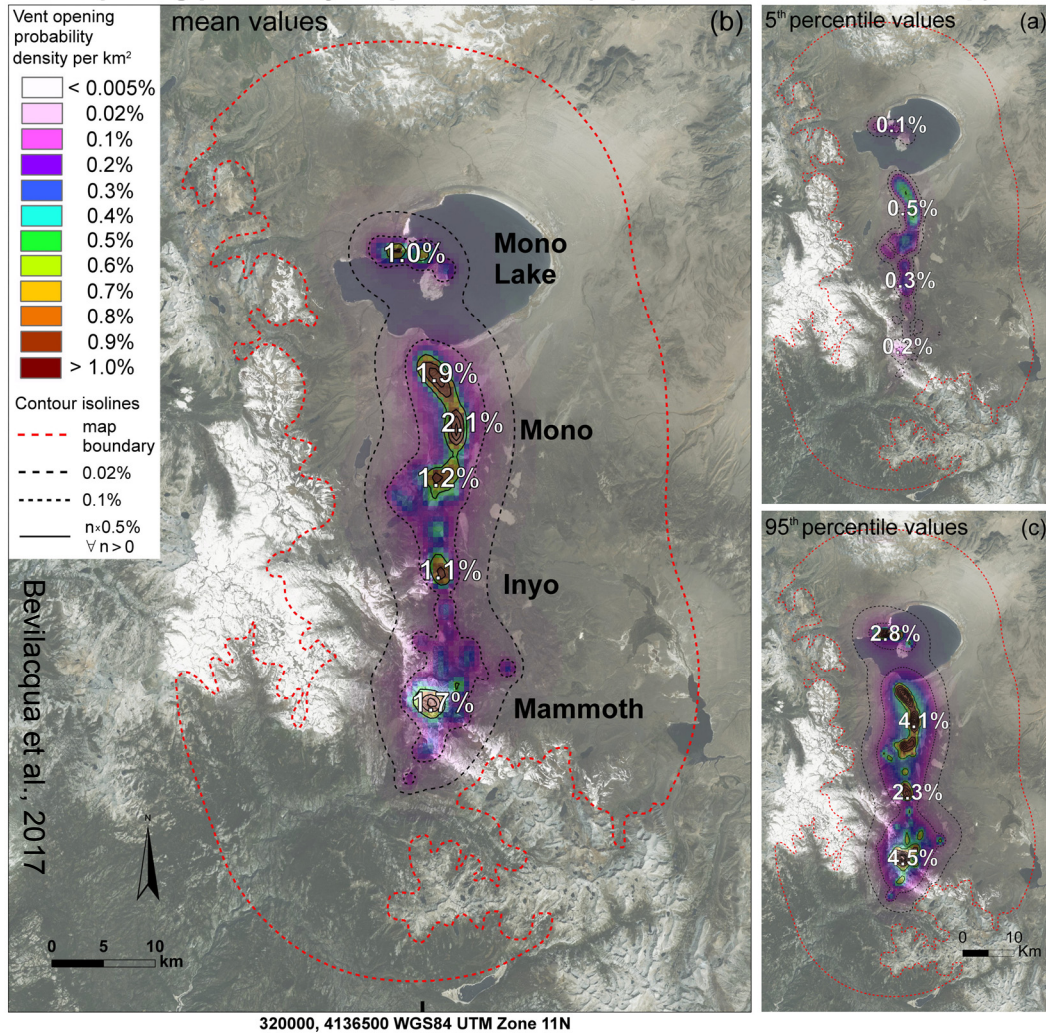
MODEL 1

A new event propagates from one **preexisting vent**, to a random distance according to a **Gaussian kernel**.

Importance of vent locations belonging to the **Mammoth Mountain** region ranges from a negligible value to equal importance with the ones in the more recently active **Mono region**.

See Bevilacqua et al., 2017 for more details on the bandwidth selection method.

Vent opening probability maps - Model 2 (Bayesian update of fault map)



- Vent opening maps -

MODEL 2

The model assumes that the new vents will likely **open near** the location of a fault outcrop ζ .

The prior probability distribution of ζ comes from **log-extension data** younger than 130 ka.

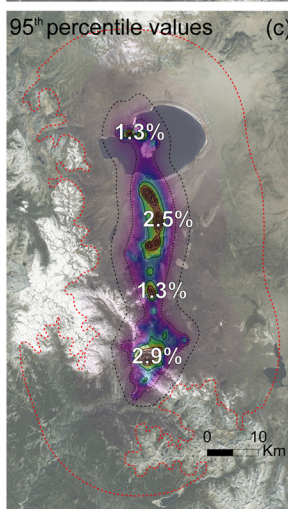
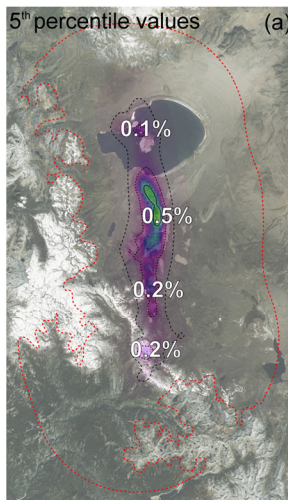
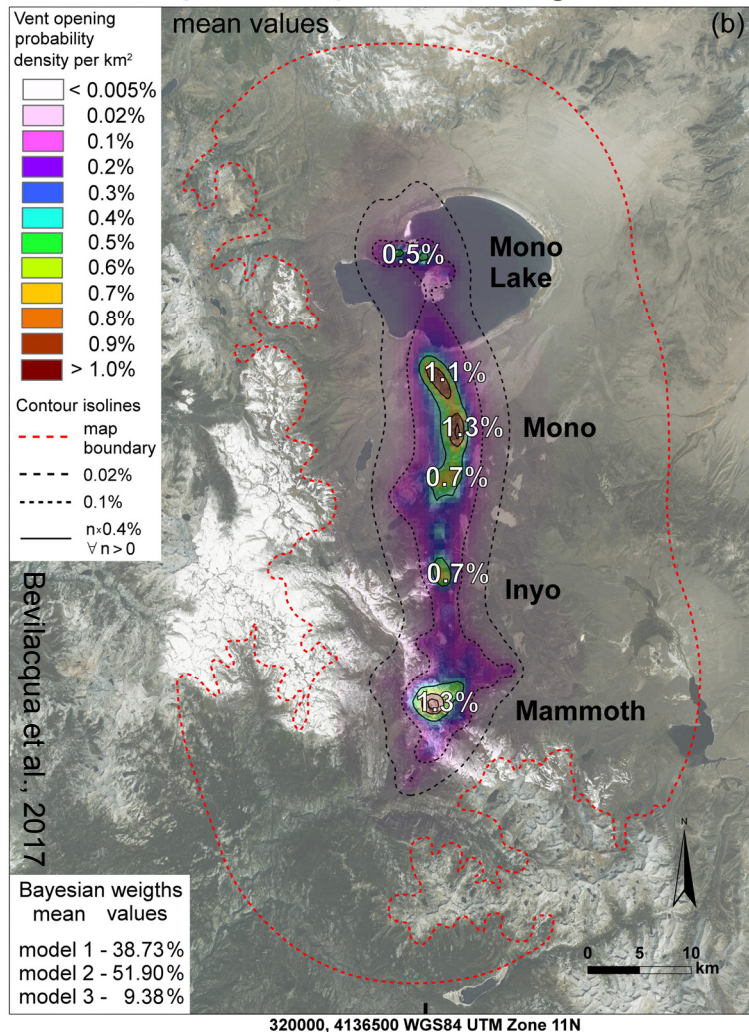
The Bayes Theorem enables us to calculate the **posterior** probability density of ζ assuming Gaussian likelihood functions around past vents locations.

It describes the fault locations that are **closer to past vents**.

The vent opening map is then obtained by **convolving** the likelihood function with the distribution of ζ .

See Bevilacqua et al. 2017 for more details, including the description of the uncertainty sources.

Vent opening probability maps - Averaged model



- Vent opening maps -

The BMA scores are proportional to the **likelihood** that the models give to the observed data (past vent sites).

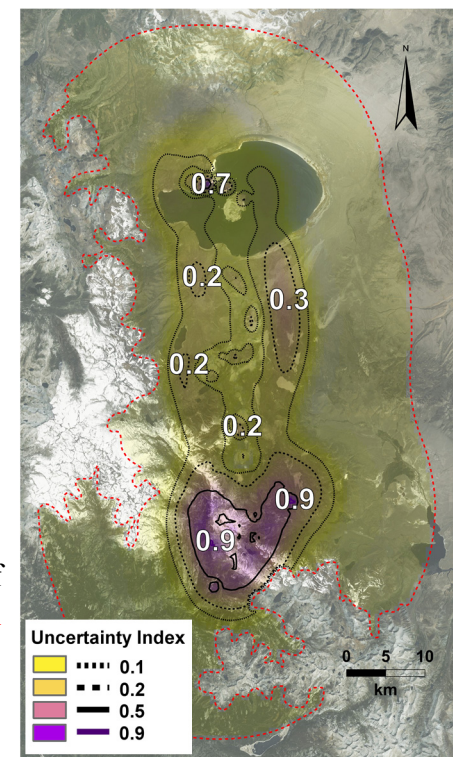
The averaged model is the **linear combination** of the different model results, with the performance scores as weighting coefficients.

These probability maps combine three models:

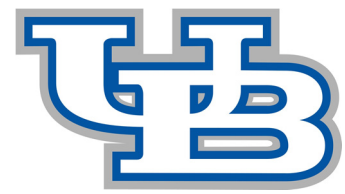
- 1 - kernel density around **past vent** locations
- 2 - Bayesian update of **fault** map
- 3 - **uniform** probability distribution on 20 km range

Figure (d) is the map of an **uncertainty index** proportional to:

$$(95^{\text{th}} \text{ perc.} - 5^{\text{th}} \text{ perc.}) / 5^{\text{th}} \text{ perc.}$$



Bevilacqua et al., 2017



- Summary and conclusions -



- A new **probability model** was developed for the effects of epistemic uncertainties affecting past eruption record.
- The procedure enabled a **statistical analysis of the temporal record** of Long Valley volcanic region.
- Three **forecasting models were compared** giving overall consistent results, with some differences.
- Two **multi-model procedures** were implemented to combine these forecasting models. BMA was also applied to the **vent opening mapping** problem, combining three diverse models.
- This study is part of a greater project aimed at the construction of a background spatio-temporal model capable of forecasting the **time and site of a future eruption** in LVVR.

