

spotaroo: spatiotemporal clustering in R of hotspot data

Di Cook award presentation

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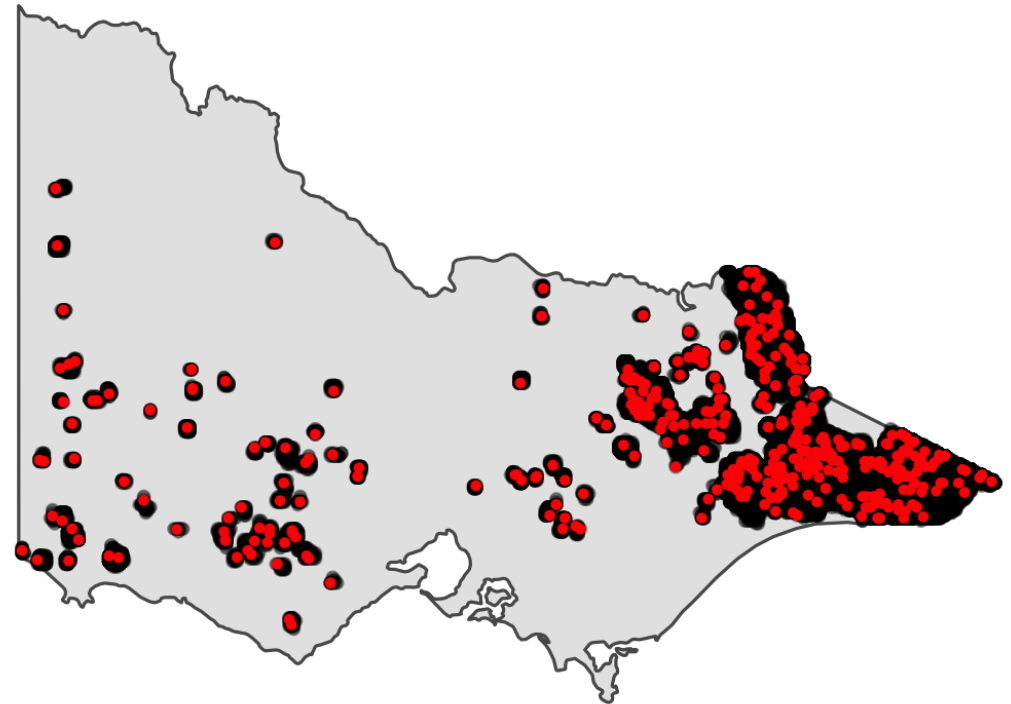
May 17, 2022

Overview of Fires and Ignition Locations

Fires Selected: 407

From: 2019-10-01 03:20:00

To: 2020-03-28 19:40:00



👮 2019-2020 Australian bushfire season 👮

A catastrophic and unprecedented disaster.

By the end of 2020,

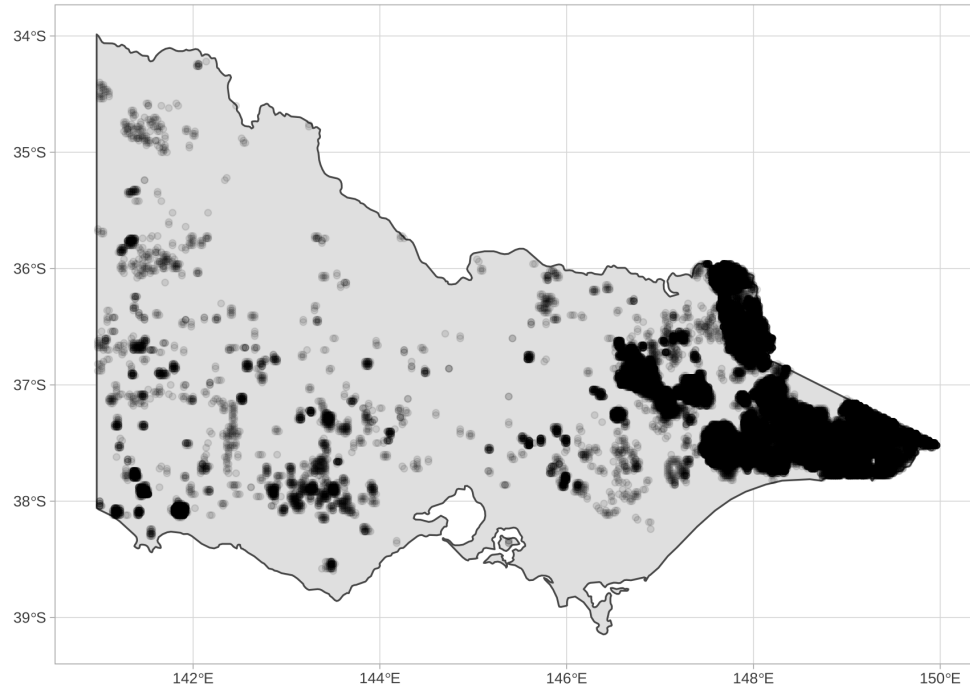
- almost **19** million hectares of land burned
- over **3000** homes destroyed
- AUD \$ **1.7** billion in insurance losses
- an estimated **1** billion animals killed



Fig. 1: An out of control bushfire in Werombi, NSW (Wikimedia Commons, 2019).



Remote sensing data



Remote satellite data provides a potential solution to the challenge of active fire detection and monitoring.

Japan Aerospace Exploration Agency (JAXA)
Himawari-8 satellite wildfire product:

- 1989572 hotspots in Australia from October 2019 to March 2020
- 0.02° (~ 2 kms) spatial grid resolution
- 10 minutes temporal resolution

Fig. 2: Himawari-8 hotspot data in Victoria from October 2019 to March 2020.

Authors of **spotoroo**

The clustering algorithm was developed in 2019, and made available in the **spotoroo** package in March, 2021.

```
install.packages("spotoroo")  
library(sporoo)
```

- Author, maintainer: **Weihaio (Patrick) Li**
- Contributor: **Prof. Di Cook**
 - Professor of Business Analytics at Monash University.
- Contributor: **Emily Dodwell**
 - Principal Inventive Scientist at AT&T.

1 Divide hotspots into intervals

activeTime: the maximum amount of time a fire may stay smoldering but undetectable by satellite before flaring up again

S_1 = hotspots observed from hour 0 to hour 1

S_2 = hotspots observed from hour 0 to hour 2

\vdots

$S_{activeTime}$ = hotspots observed from hour 0 to hour $activeTime$

$S_{activeTime+1}$ = hotspots observed from hour 1 to hour $activeTime + 1$

\vdots

S_T = hotspots observed from hour $T - activeTime$ to hour T

In summary, data will be divided into T intervals

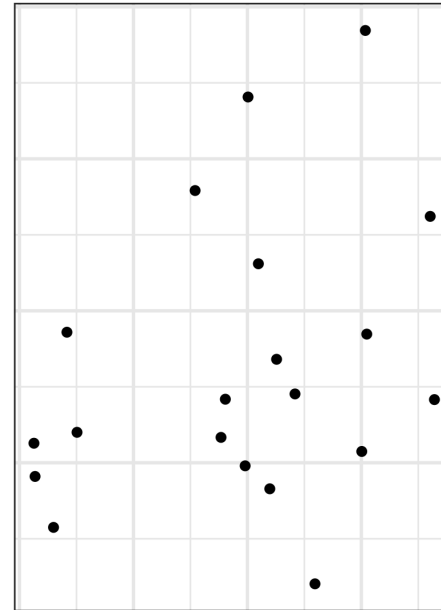
$$S_t = [\max(1, t - activeTime), t] \quad t = 1, 2, \dots, T.$$

2 Cluster hotspots spatially within each time interval

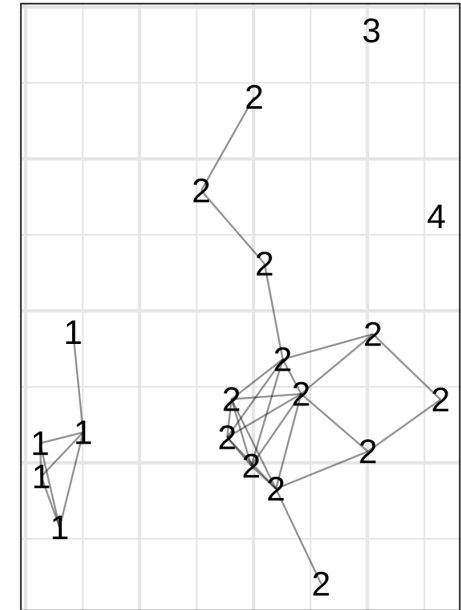
For each time interval S_t , connect all hotspots within **adjDist** to form a graph

Every **connected component** of the graph is an individual cluster

S_t before clustering



S_t after clustering

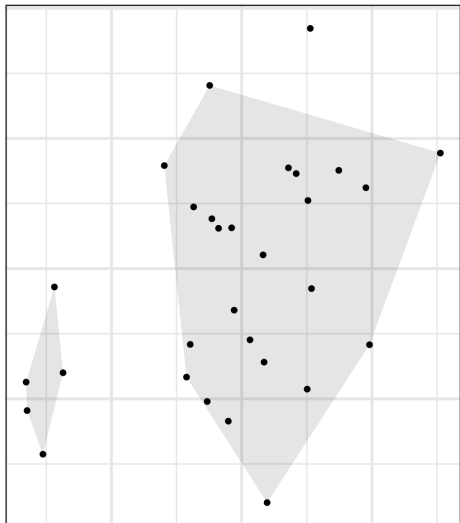


3 Update memberships for hotspots

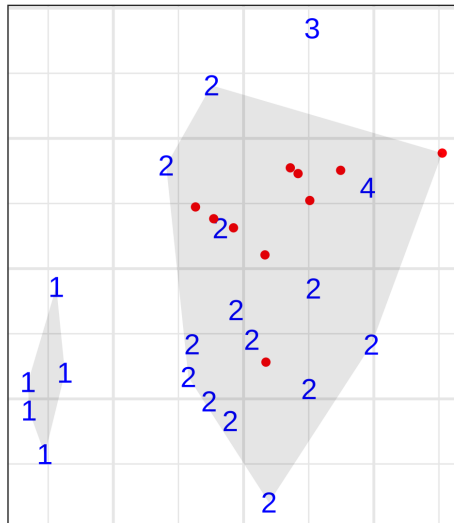
Some hotspots in S_t have been clustered in S_{t-1}

Update the membership by finding the nearest label

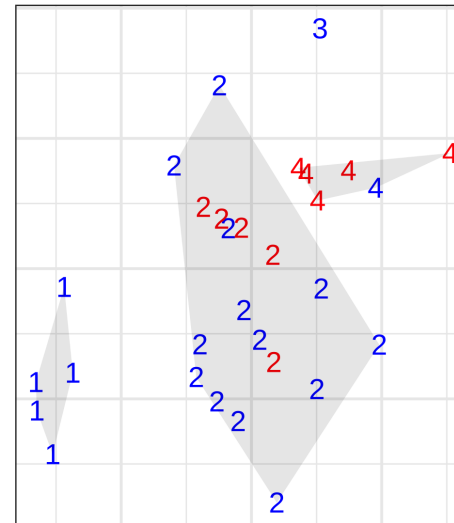
S_t before updating



Adopt information from S_{t-1}



S_t after updating



4 Handle noise in the clustering result

- `minPts`: the minimum number of hotspots in a cluster
- `minTime`: the minimum time a cluster can exist and still be considered a bushfire

Any cluster that does not satisfy these two conditions will be reassigned membership label -1 to indicate they represent noise.



Usage

```
result <- hotspot_cluster(hotspots,  
                          activeTime = 24,  
                          adjDist = 3000,  
                          minPts = 4,  
                          minTime = 3,  
                          timeUnit = "h",  
                          timeStep = 1)
```

SPOTOROO 0.1.2

— Calling Core Function : `hotspot_cluster()` —

— "1" time index = 1 hour

✓ Transform observed time → time indexes

i 4313 time indexes found

— activeTime = 24 time indexes | adjDist = 3000 meters

✓ Cluster

i 1055 clusters found (including noise)

— minPts = 4 hot spots | minTime = 3 time indexes

✓ Handle noise



Summary of the clustering result

```
summary(result)
```

SPOTOROO 0.1.2

— Calling Core Function : `summary_spotoroo()` —

CLUSTERS: ALL

OBSERVATIONS: 75936

FROM: 2019-10-01 03:20:00

TO: 2020-03-28 19:40:00

— Clusters

i Number of clusters: 407

Observations in cluster

Min.	1st Qu.	Mean	3rd Qu.	Max.
4.0	17.0	178.5	159.0	3863.0

Duration of cluster (hours)

Min.	1st Qu.	Mean	3rd Qu.	Max.
2.3	9.7	43.4	63.4	285.5



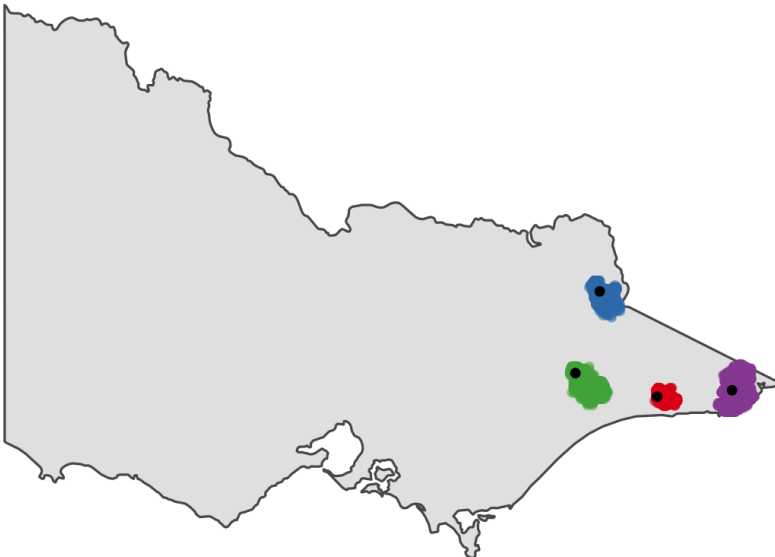
Visualizing the clustering result

Spatial distribution of clusters

```
plot(result, bg = plot_vic_map(),  
      cluster = c(58, 83, 129, 163))
```

Overview of Fires and Ignition Locations

Fires Selected: 4
From: 2019-10-01 03:20:00
To: 2020-03-28 08:30:00

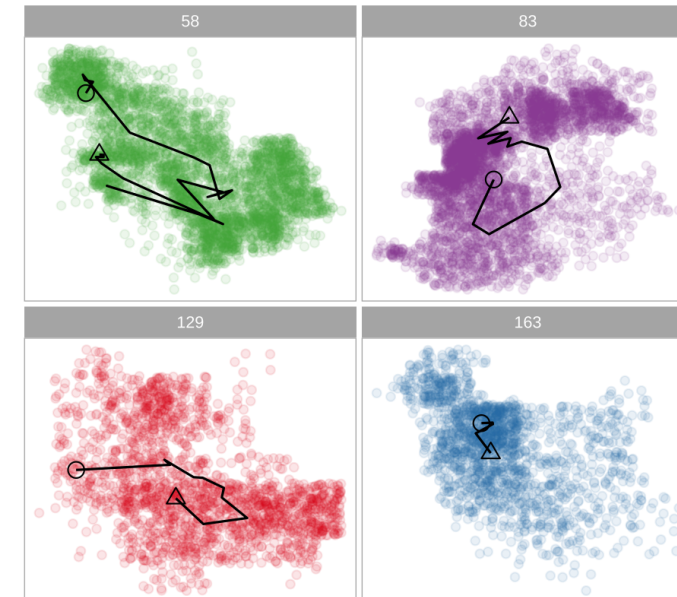


Fire movement plot

```
plot(result, type = "mov", step = 12,  
      cluster = c(58, 83, 129, 163))
```

Fire Movement (Δ : Start | O: End)

Fires Selected: 4
From: 2019-12-18 14:30:00
To: 2020-01-04 18:50:00

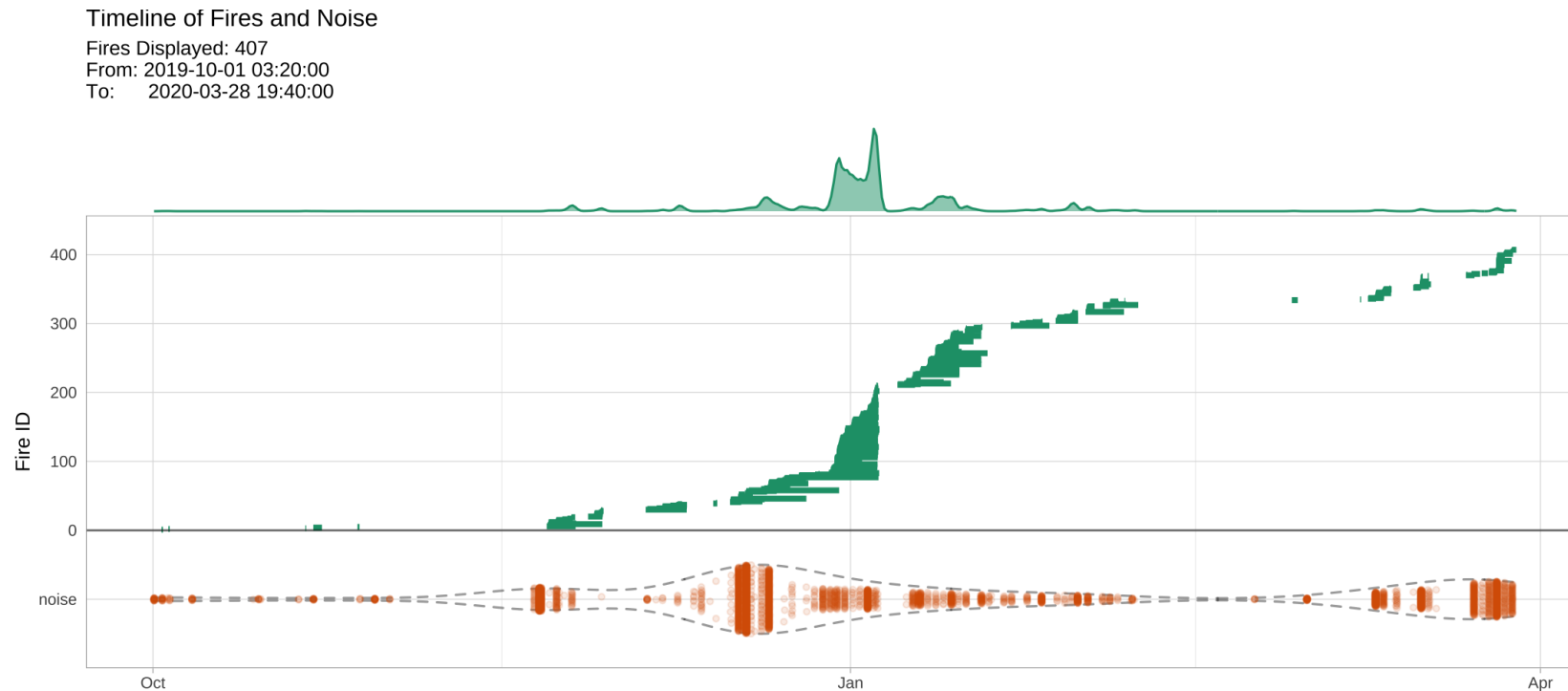




Visualizing the clustering result

Timeline plot

```
plot(result, type = "timeline")
```



Thanks!

Bibliography

Wikimedia Commons (2019). "Werombi Bushfire".

https://commons.wikimedia.org/wiki/File:Werombi_Bushfire.jpg

Li, W., Dodwell, E., & Cook, D. (2021). A Clustering Algorithm to Organize Satellite Hotspot Data for the Purpose of Tracking Bushfires Remotely. <https://github.com/TengMCing/Hotspots-Clustering-Algorithm/blob/master/li-dodwell-cook/RJwrapper.pdf>