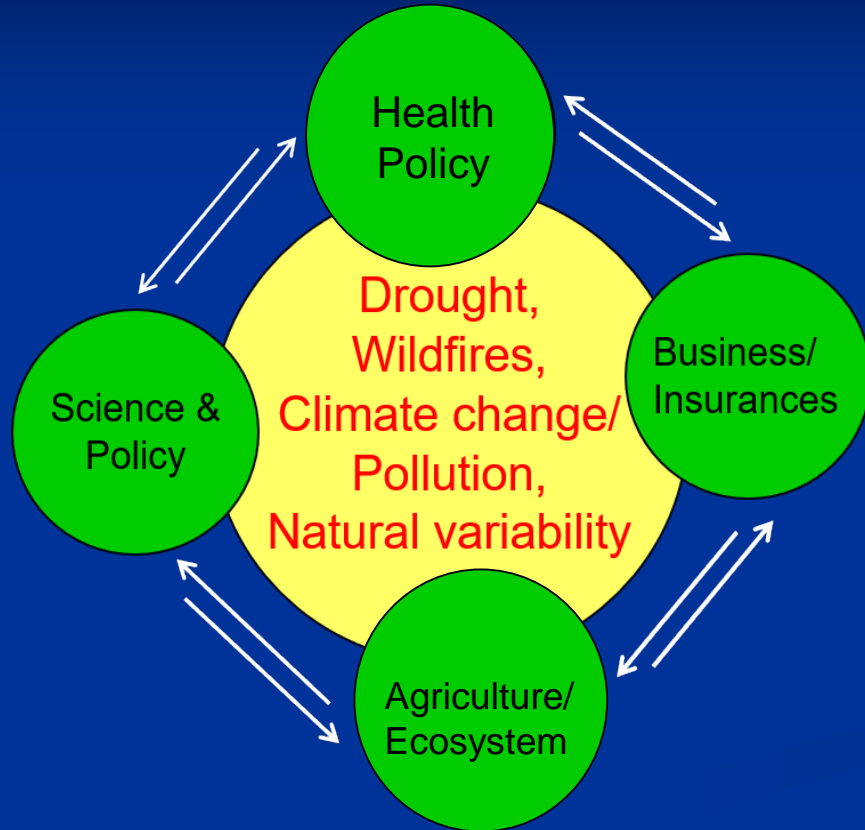


Climate Change and Wildfires in California

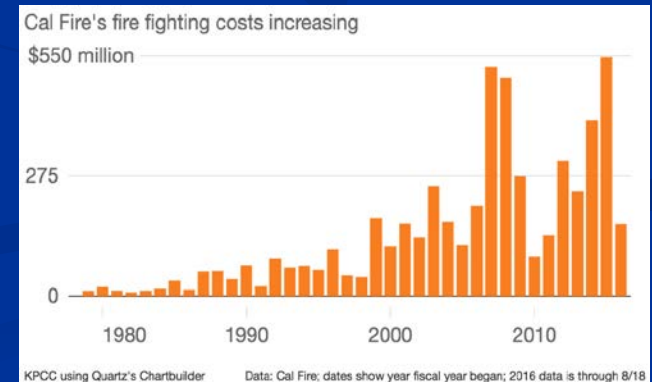
Understanding Complex System Interactions using Bigdata and ML



Kids health matter



Current cost ~ \$550 million in California, tripled over the last decade.



Ozone a strong oxidant , produce free radicals in the body

O₃ and PM_{2.5} can enter Respiratory system a



Study area

San Diego, CA

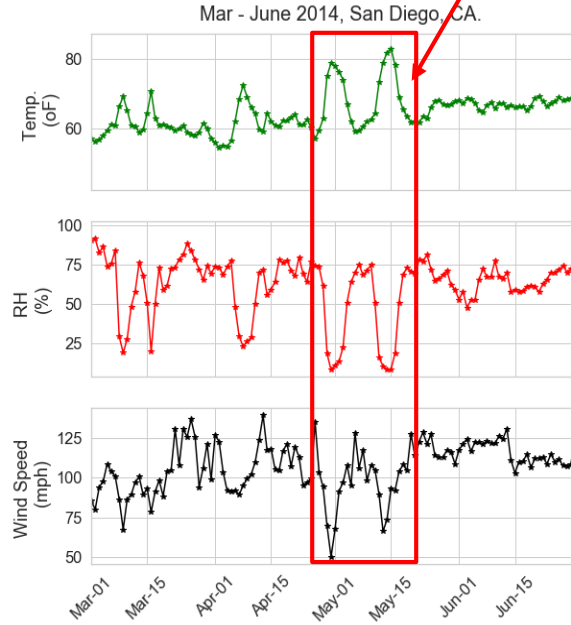


GOAL1: Atmosphere-Biosphere Linkage to predict Wild Fires

Previous Work

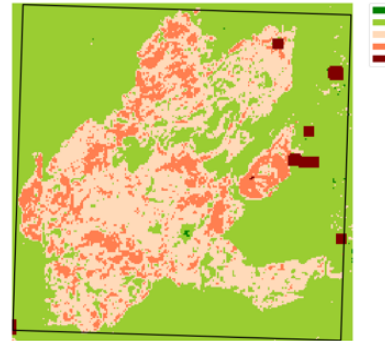
High Temperature
Low RH
Windspeed

Figure 1(a-e). Atmospheric conditions during wildfires



Multispectral Remote Sensing Data, Landsat-8 , DNBR Difference in Normalized Burn Ratio

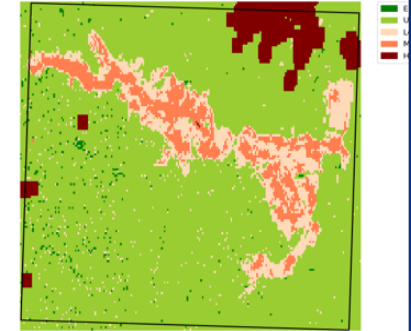
Fig. 10. Landsat Derived dNBR 09 June vs 25 May 2014
Tomahawk Fire, San Diego, CA.



Tomahawk Fire,
Burned area calculation:
Class 4,5 ~ 5 km²
Class 4,5,6 ~ 91 Km²

Reported CalFire = 70Km²

Fig. 9. Landsat Derived dNBR n 09 June vs 25 May 2014
Bernardo Fire San Diego, CA.

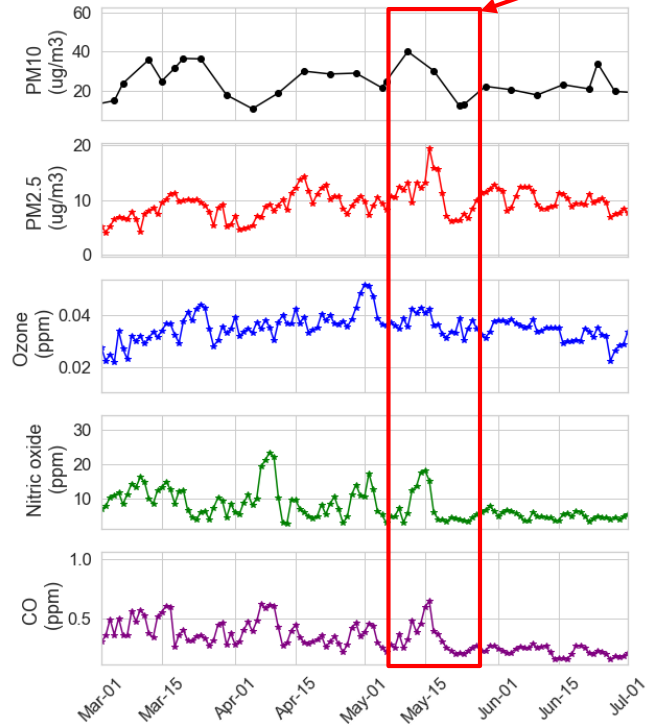


Bernardo Fire
Burned area calculation:
Class 4,5 ~ 3.1Km²
Class 4,5,6 ~ 9.5 Km²

Reported CalFire = 6.3Km²

Figure 2(a-e). Chemical composition of the Atmosphere during wildfires

Mar-June 2014, San Diego, CA.



2014 Wildfires in San Diego = Increase in Particulate matter and toxic gases in the atmosphere

Burned 110 Km²
Cost ~ 90 million USD +
40 million property damages +
10 million health related cost



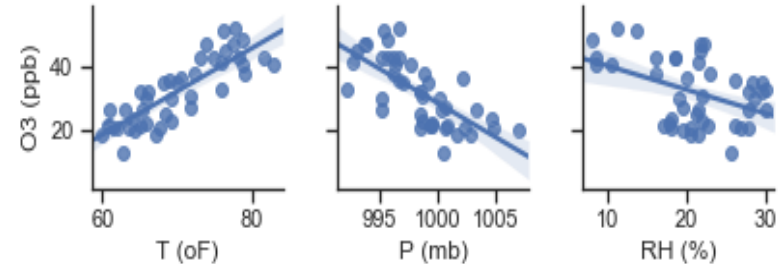
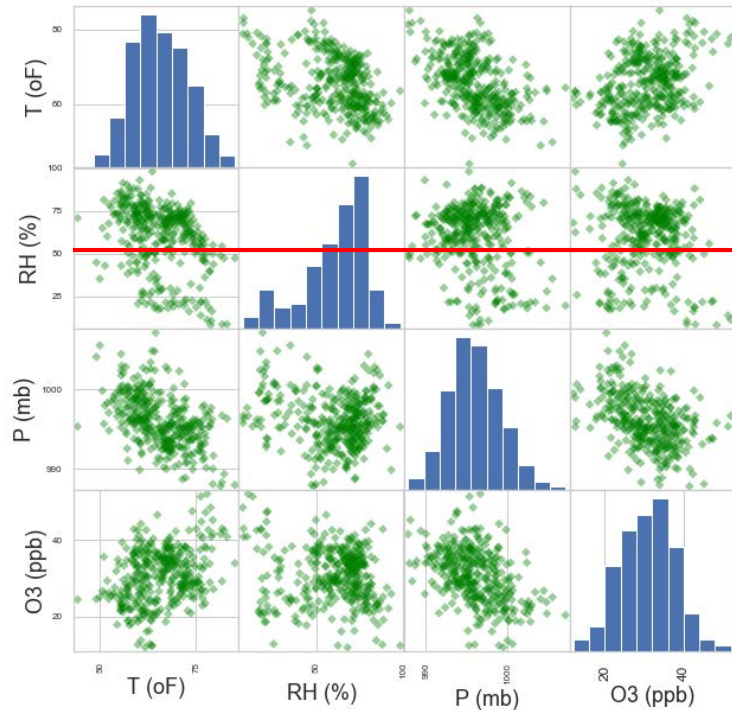
GOAL2: Machine Learning to predict Fires and Pollutant Concentrations

Correlation matrix

Salient Features

1. Interaction between Temperature and Relative humidity is complex in San Diego showing two distinct populations.
2. Ozone formation is temperature and pressure dependent.
3. RH and ozone concentration requires further analysis ($50\% > RH < 50\%$)

RH < 30% = Santa Ana Days



Interaction between atmospheric conditions with ozone concentrations.

Statistical Analysis with Scikit learn (single variable)

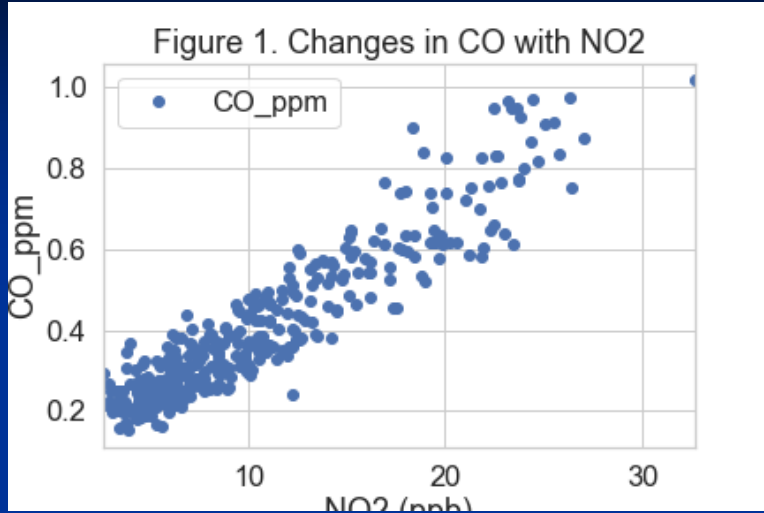


Figure 2. Distribution pattern of NO2

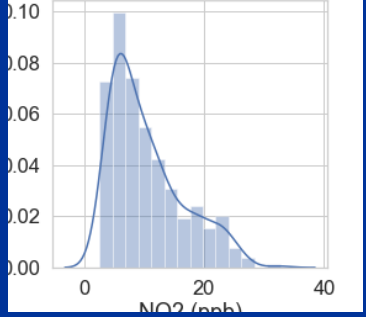
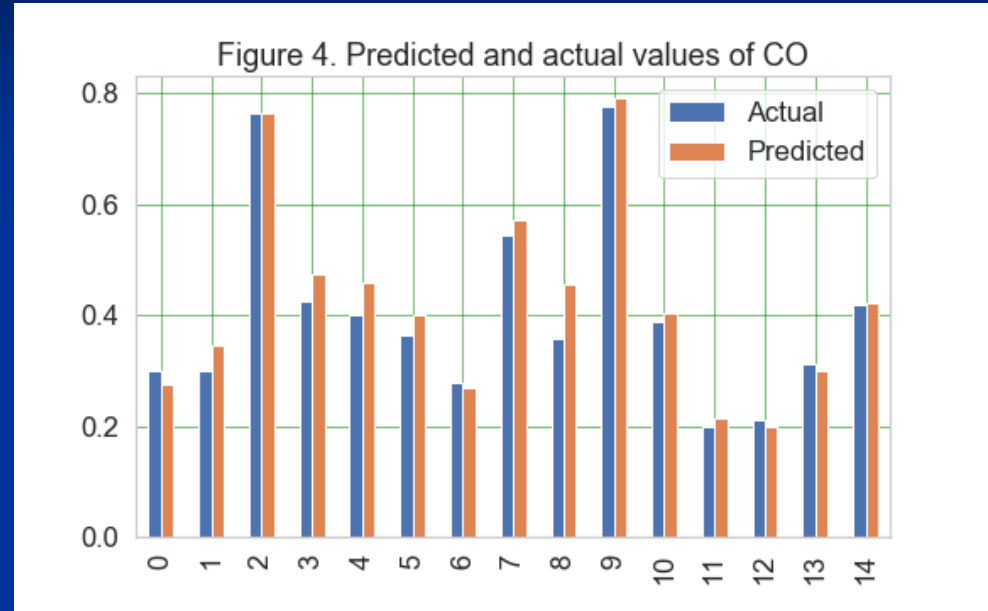
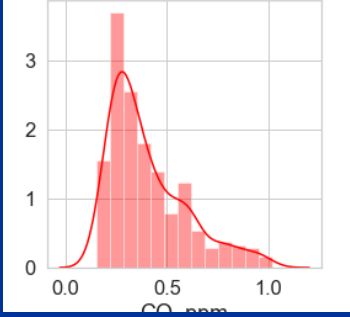


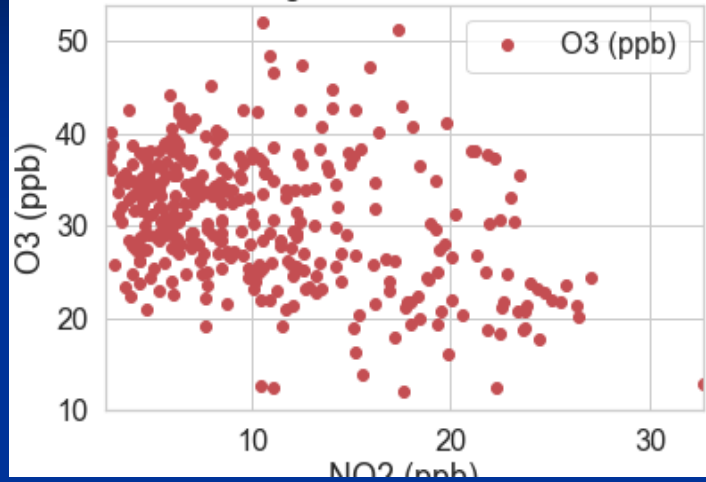
Figure 3. Distribution pattern of CO



Statistical Analysis with Scikit learn (multivariable analysis)

Ozone, with (NO_2 , CO, $\text{PM}_{2.5}$)

Figure 6. O₃ vs NO₂



Comparison of actual and predicted values of ozone

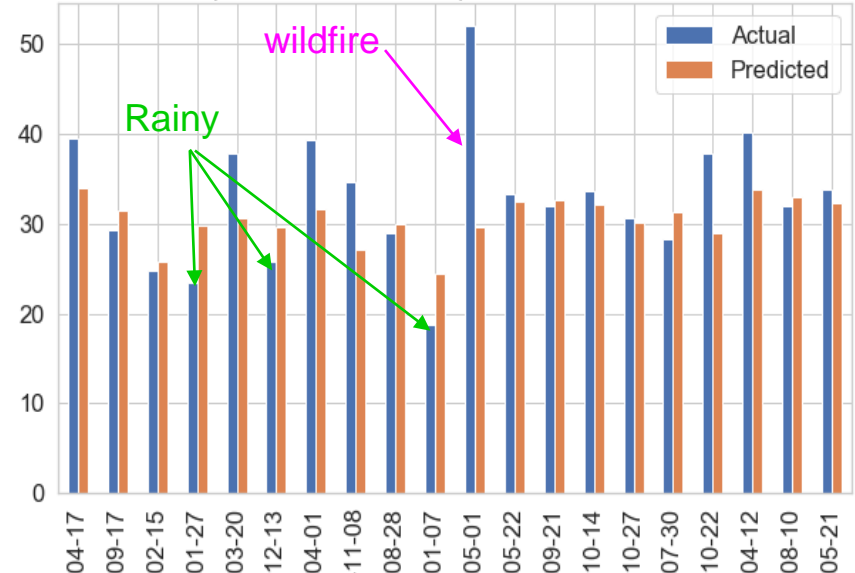


Figure 2. Distribution pattern of NO₂

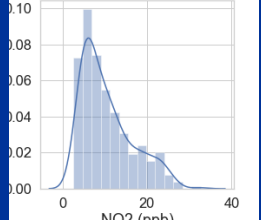
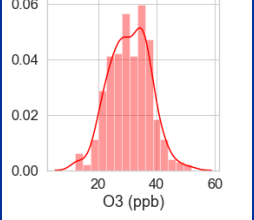
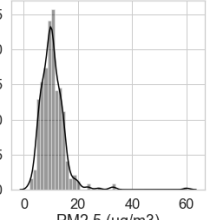


Figure 5. Distribution pattern of O₃



Distribution pattern of PM_{2.5}



R values=

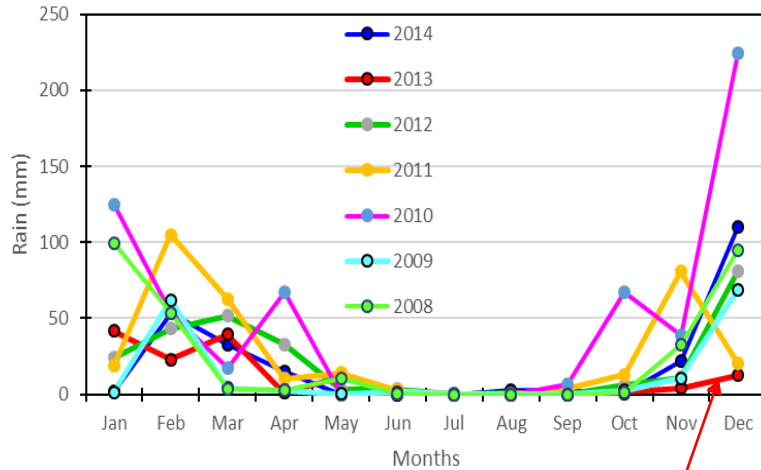
O₃ vs NO₂ = -0.445

O₃ vs PM_{2.5} = -0.792

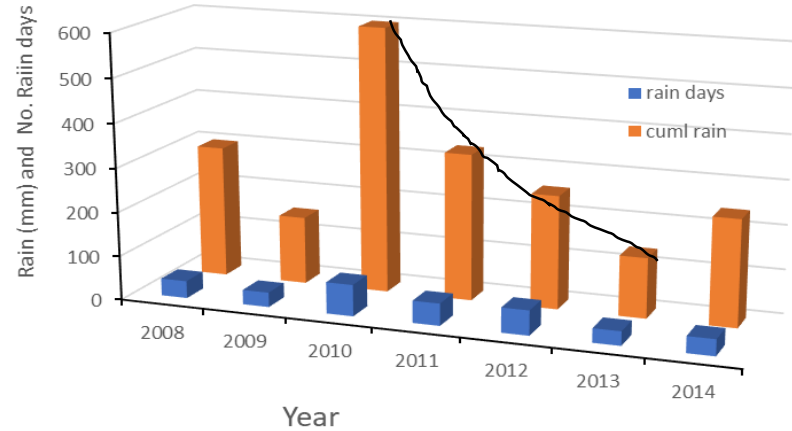
O₃ vs CO = 0.2025

Link between Rains in San Diego and Wildfires

Monthly Rain in San Diego 2014



Cumulative Rain (mm) and Rain Days, San Diego, CA. 2008-14

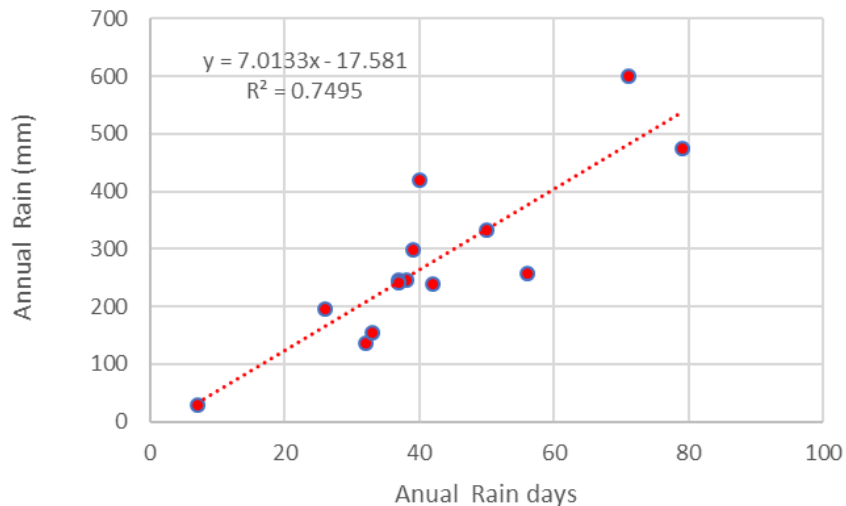


December rain is critical to prevent wildfires in San Diego

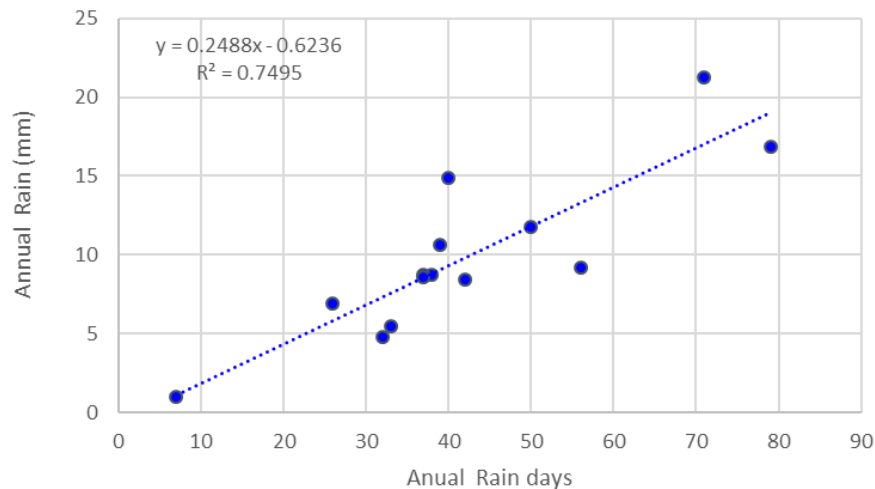
Lessons Learned about Data Pipelining

Treating data points like molecules and using Thermodynamics and kinetics to find order of reactions can serve us best

Annual Rain (mm) vs rain days



(A/A_{\min}) (mm) vs rain days



Summary and Future Work

1. We can predict wildfires using atmospheric data and Landsat IR bands
2. Potential to save lives and properties and health.
3. Reduce cost to combat wildfires.
4. Huge potential to refine Algorithm using Bigdata (data munging and pipeline to reduce errors) to predict and prevent wildfires.
5. One shoe does not fit all, we have to apply site specific models (coastal or inland)