



GWFF

GEOSPATIAL WORLD FORUM

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ForestRe

GLOBE
UNDERWRITING

Utilising geo spatial data - risk assessment & pricing

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Founder & MD ForestRe

&

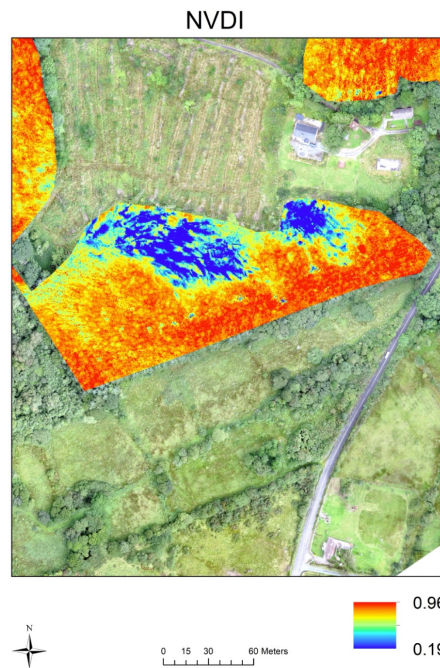
Yuming Zhi:

Principal Analyst & Underwriter at Globe

Geospatial World Forum (GWF) 2024

“Geospatial Transition:
Powering the World Economy”.

Rotterdam, the Netherlands; 14 – 16 May 2024



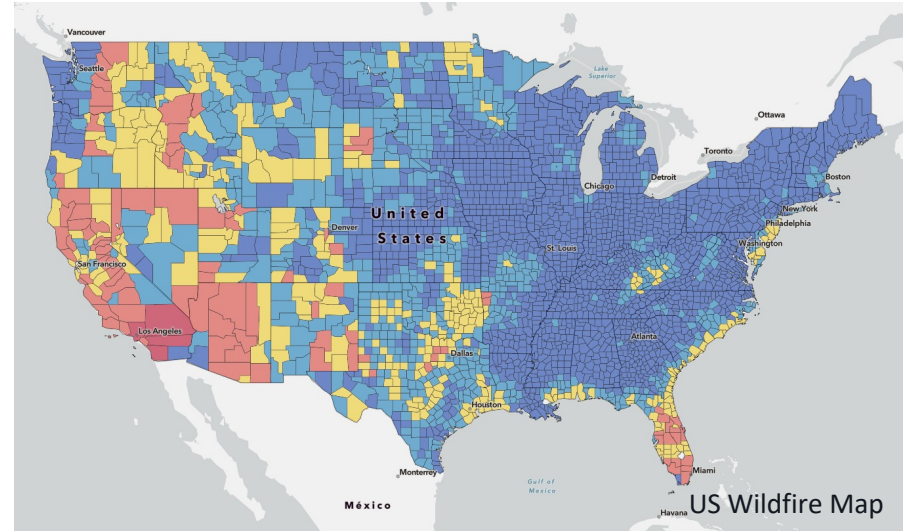
Source; 2017 by-



AERIALAGRITeCH

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- Introduction – who are we & what we do?
- Cat. Risks in forestry
- What do we measure?
- How do we analyse the risk?
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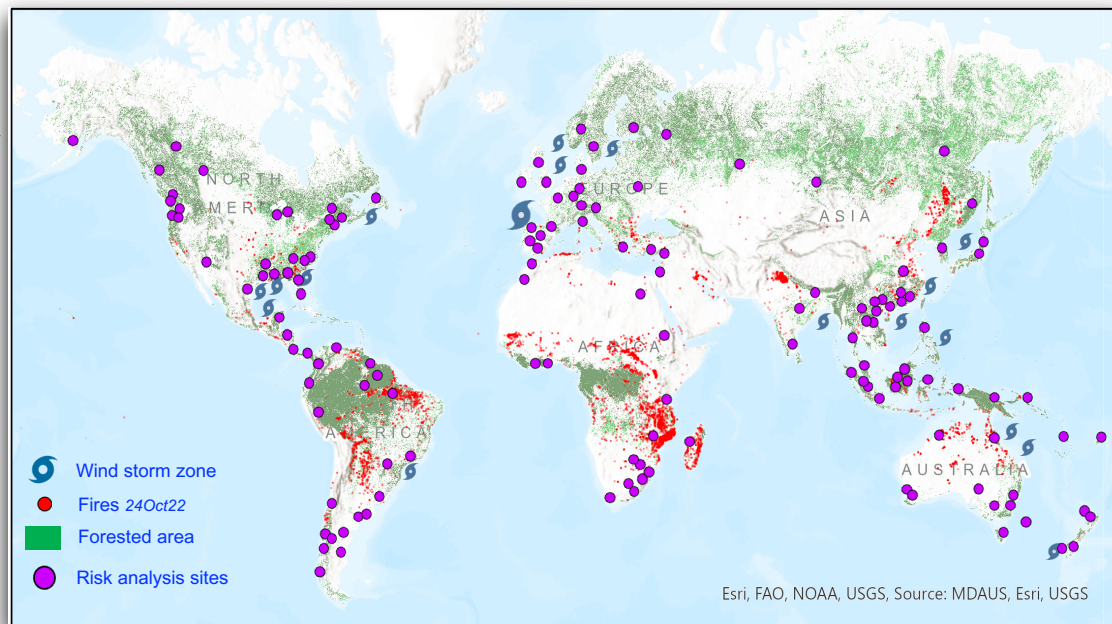
Source: <https://hazards.fema.gov/nri/wildfire> at 11 May 2024

Worldwide risk profiling

- **'ForestRe' (FRe)** is a technical team with 30-years experience in forestry risk profiling, pricing & management
- The team has developed **the tools using geo data, supported by extensive global experience** to quantify those financial risks...
- Across **forestry, agro-forestry and carbon investments**

Global reach of ForestRe

Map of our forest risk analysis sites



- Fully FCA authorised Lloyd's cover holder
- Insure in 30 countries / 6 continents
- Supported by our large panel of global insurers

Catastrophic hazards in forestry

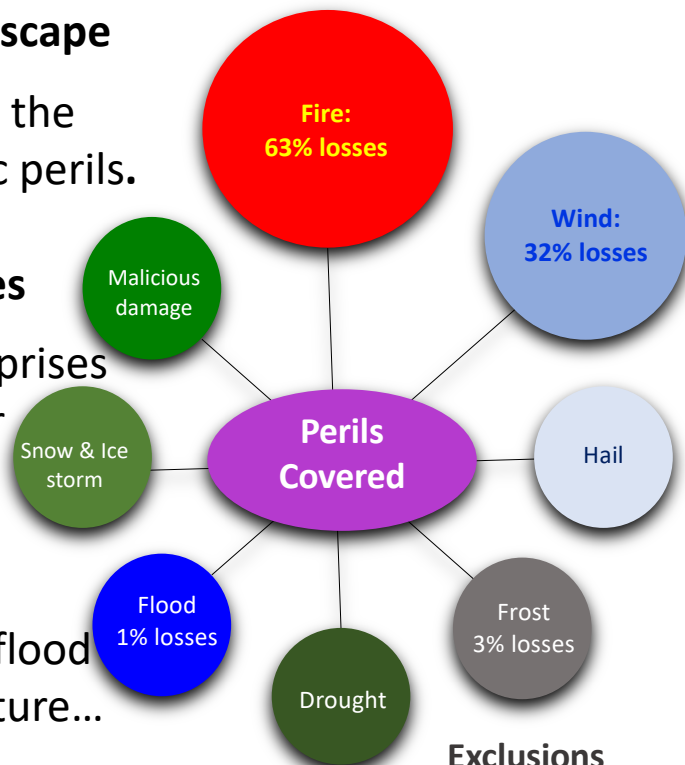
Forestry is a landscape

Fire and wind are the main catastrophic perils.

Fruit trees & vines

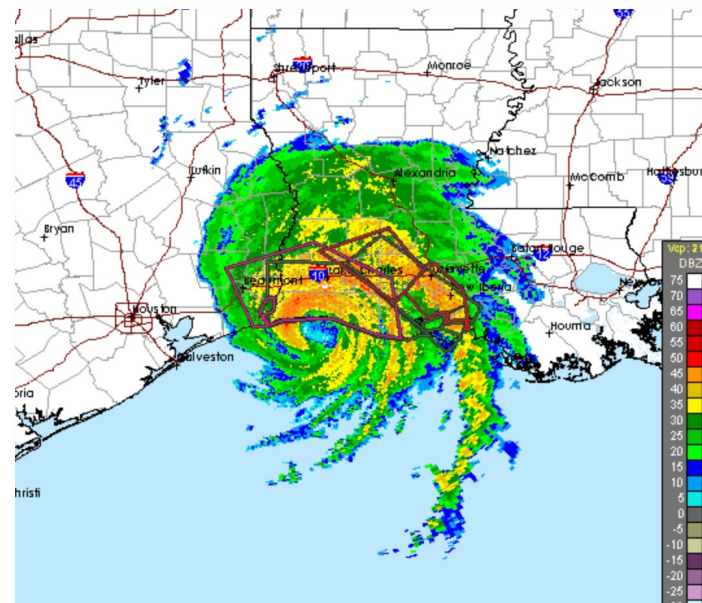
These local enterprises have a far greater exposure to local events:

Hail, wind, frost, flood & excessive moisture...



Exclusions

- Subterranean Fire
- Pests & disease



Hurricane Laura near peak intensity while approaching Louisiana late on August 26 2020; Highest winds 150 mph (Cat. 4):

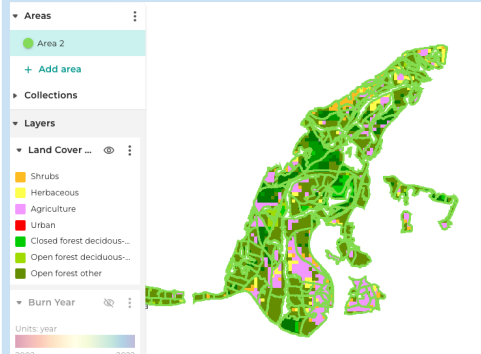
Source: https://en.wikipedia.org/wiki/Hurricane_Laura

- What do we measure with Geo-spatial data ?

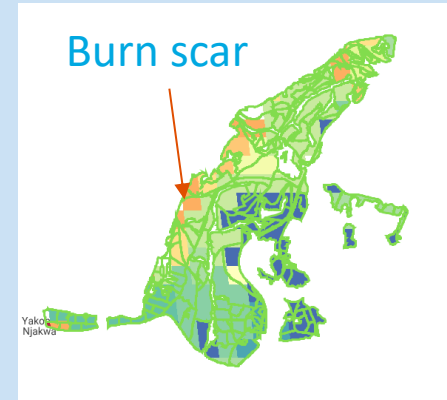
Area of the plantation



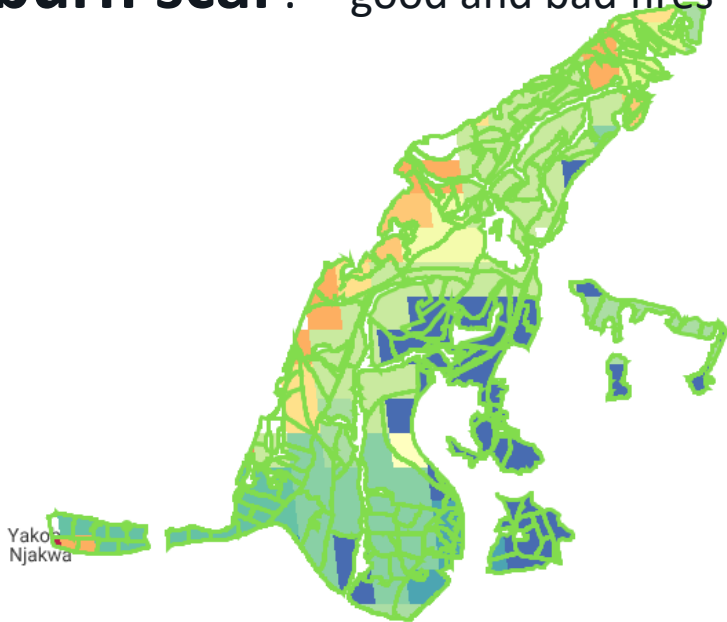
Landcover Type



Historical Burn Scar



How to interpret historical burn scar? – good and bad fires



Our aim is to measure fires that damaged the value of the forest since 2001

From that to determine:

- Severity
- Frequency

Not all fires are ‘bad’

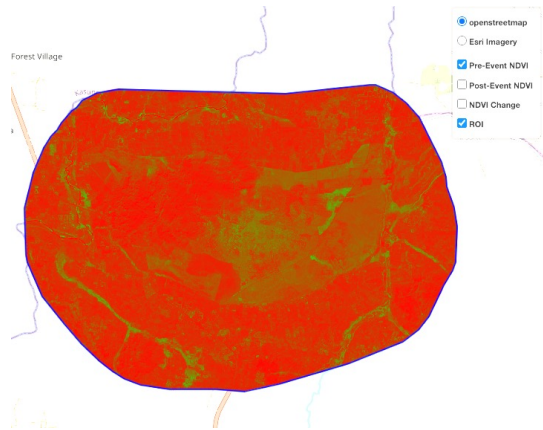
- Some fire prevent bigger fires – ‘good’ fires
such as fuel reduction – burning off ground fuel such as weeds, shrubs, invasive tree species
- Burning along fire breaks so fires do not pass.

We need good geo-spatial data to help us do this.

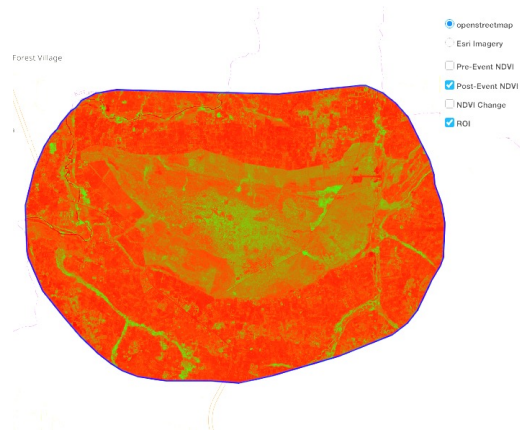
Damaging fires - NVDI

Good Fire (fire to reduce fuel load to prevent bigger fires in future) or Bad Fires?

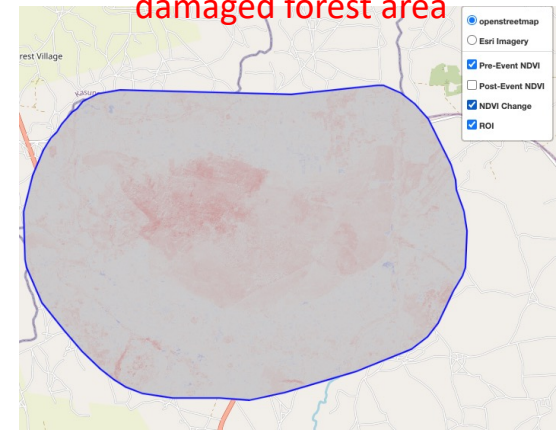
Before event



After event



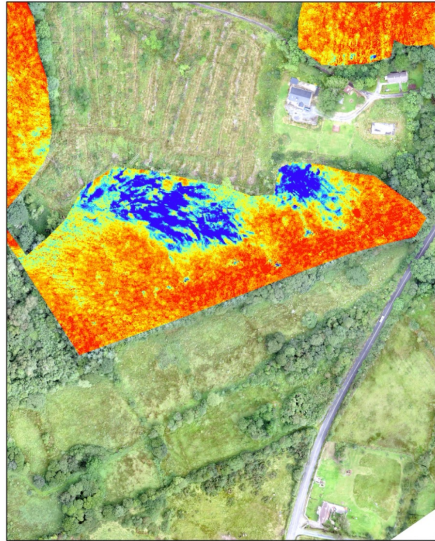
The difference being damaged forest area



*NDVI Justification: reflects the reduction in live vegetation and the exposure of charred surfaces, compared to the pre-fire measurements that indicated healthier and denser vegetation cover.

Wind damage – NVDI also useful

NVDI



NDVI Justification:

reflects the reduction in live vegetation and the exposure of wind-blown trees, compared to the pre-storm measurements

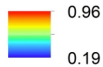
Essential for us to help acquire empirical wind damage data for any client.

Other tools:

Surface depression of forest canopy due to wind.



0 15 30 60 Meters



Source; 2017 by-



How do we analyse the risk?

Shapefile Validation

Key: Is the kmz file complete and accurate?

Workflow Creation

Find the right dataset and set the correct parameters

Data Analysis & Input

Landcover Type Detection
 Burn Scar Mapping and Burn Data Validation
 Landcover and Burn Scar Mapping

Pricing Model

Loss cost calculation:-
 Monte Carlos Simulation for Loss iteration
 Loss Distribution Fitness Check
 Generating Technical Pricing



Risk Assessment Report

Item Insured (USD)	56,789,587.88	56,789,587.88
Value at Risk		
Total Area (Ha)	34,622	34,622
	per hectare	1,640.3
Limit (USD)		
FLEXA Each and Every Loss	7,000,000	7,000,000
FLEXA Annual and Aggregate Loss	7,000,000	7,000,000
Total Aggregate Limit	7,000,000	7,000,000
Sublimit		
Fire Fighting Cost	250,000	250,000
Defectibles (USD)		
Cover A FLEXA		
Each and every loss	20,000.00	20,000.00

Indication Quote

What geo spatial data do we use?

Land Cover Data

- Copernicus Global Land Cover (2019)
- Dynamic World Landcover (10m)
- Planet – NICFI tropical regions

Burn Scar Data

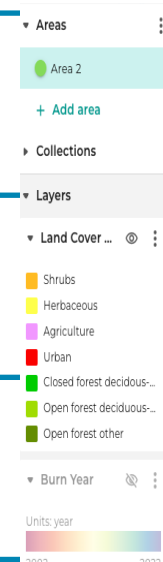
- FIRMS: Fire Information for Resource Management System (1km)
- Modis Burned Area Monthly Global (500m)

Land Image Data

- Landsat 4/5/7/8/9 Combined (30km)
- Sentinel-2 MultiSpectral (10-60m) - NDVI

Weather Data

- ERA5 Land Daily Aggregated Reanalysis
- ERA5 Land Monthly Aggregated Reanalysis (11km)



What platform do we use?

- 1) Collaborated since 2019
- 2) This data feeds into our **In-house pricing tool** through API connecting to EarthBlox for data

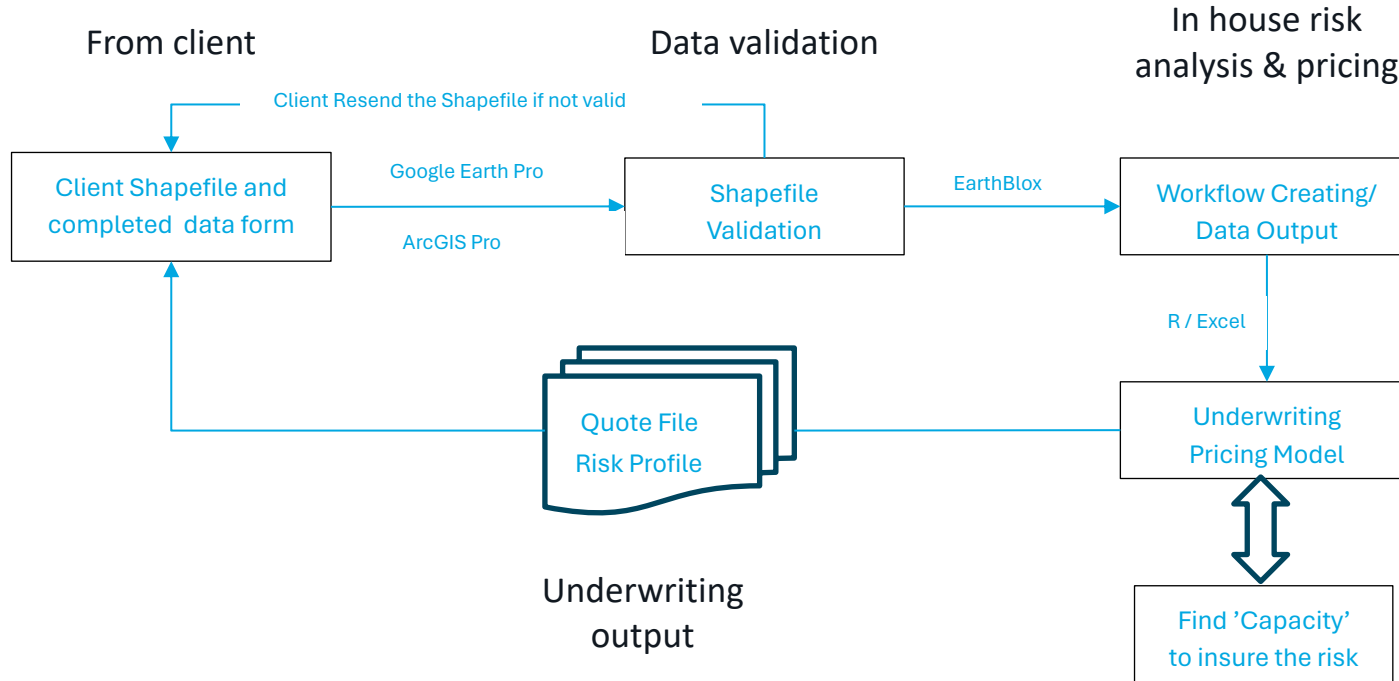


EARTH BLOX



Source; Earthblox: Land cover change data
<https://www.earthblox.io/packages/land-cover-change>

Case Study – Pricing Process





Case Study – Malawi Account

1. Parameters building

Add Table

Define values

Coverage calculation
Coverage (hectares)

+ Add value

Group table

Dataset	Image band	Dates
Copernicus Global Land	discrete_classification[1]	2019

+ Add group

Data table output setting

Description

Lego like
Drag, Combine & Run

Use this Dataset
MODIS Burned Area Monthly Global (500m) ← **Dataset**

Select Area of Interest
Area of Interest: Area 2 ← **Location**

Select Time Period
From: 2001-01-01 To: 2022-12-31

Aggregate Images in Time
Combine images By year
Creating a new image every 1 year
Using the Most Recent

Classify Image Using Thresholds

Add Table
Name: Area Burnt Per Year ← **Data Layer**

Save Data for Re-use
Name: Burn

Aggregate Images in Time
Combine images Over entire time period
Using the Most Recent

Add Map Layer
Layer name: Burn Year

Use this Dataset
Copernicus Global Land Cover (100m) ← **Dataset**

Select Area of Interest
Area of Interest: Area 2

Select Time Period
From: 2019 To: 2019

Add Map Layer
Layer name: Land Cover Classification
Select single band
discrete_classification

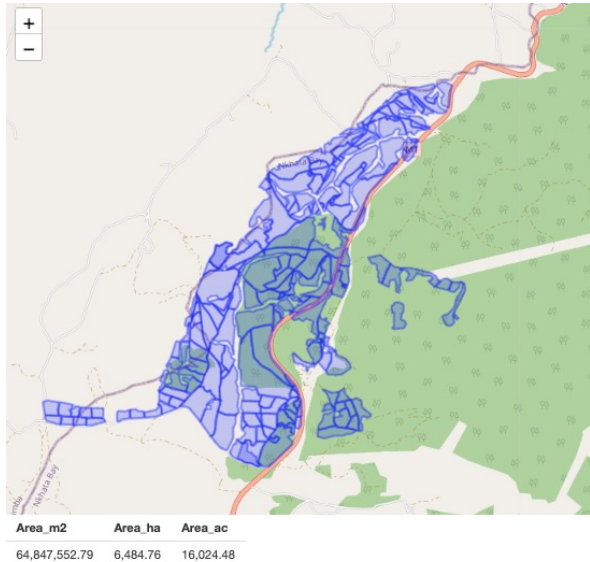
Add Table
Name: Land Cover Area (Ha)

Add Table
Name: Burn Area per Land Cover (Ha)

Stop

Case Study – Malawi Account

2. Area Check

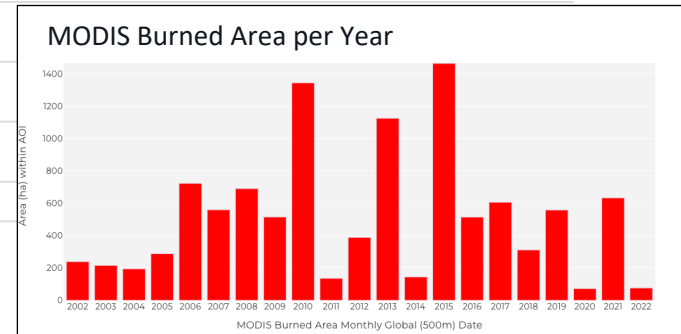


<5% Discrepancy - schedule and KMZ areas

*Assuming the shapefile has been validated

3. Landcover Type

Land Type	Area (ha)	Area (ac)	Percentage
Shrubs	868.7	2,146.6	13.40%
Herbaceous	376.5	930.5	5.81%
Agriculture	1,536.0	3,795.5	23.69%
Urban	2.4	5.9	0.04%
Closed forest evergreen-broad	70.3	173.7	1.08%
Closed forest deciduous-broad	57.2	141.3	0.88%
Closed forest other			
Open forest evergreen-broad			
Open forest deciduous-broad			
Open forest other			
Total			

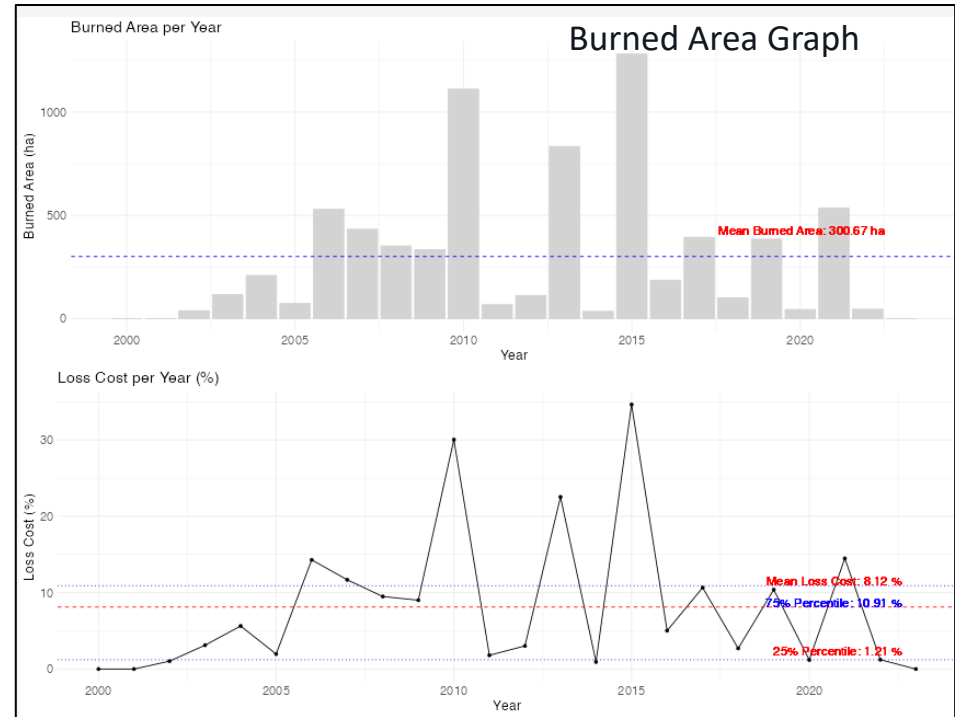


Case Study – Malawi Account

3. Generating ‘Loss Cost’ Table

	Year	Burned Area (ha)	Loss Cost
1	2000	0	0.00%
2	2001	0	0.00%
3	2002	38	1.03%
4	2003	116	3.12%
5	2004	209	5.63%
15	2014	35	0.93%
16	2015	1,282	34.64%
17	2016	186	5.02%
18	2017	394	10.66%
19	2018	100	2.70%
20	2019	385	10.40%
21	2020	44	1.19%
22	2021	536	14.49%
23	2022	45	1.22%
24	2023	0	0.00%

‘Loss cost’ (=area burnt/total area as %)



Case Study – Parametrics in Brief

Provides payouts based on pre-set parameters linked to an event, rather than traditional loss assessment.

Trigger Events

Payouts are triggered by measurable events such as wind speed, earthquake magnitude, or rainfall levels.

Rapid Settlements

Benefits include quicker claims processing and payouts due to the use of objective data.

Transparency

Clear terms and pre-defined triggers enhance transparency between insurers and policyholders

Innovation in Pricing

Offers opportunities for innovative pricing models in emerging risks like climate change.

This is the structure we have developed:

One-Stop-Data supplier

Powerful Data Processing

Real-Time
Data Access

API access

Scalability

Collaboration
Features



Source: ChatGPT 4

Summary

1. The 'ForestRe' team introduced geospatial data into all routine underwriting in 2019
2. ...To measure loss frequency and severity especially for fire
3. The geo data platform supplemented the empirical data from clients
4. Geo data provides an ability to underwrite globally without any location limits
5. BUT great care required to interrogate the data – what is burning; was there an economic loss?
6. Work in progress to develop tools for accurate long term wind events
7. We still have lots to learn...

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