



Artificial Intelligence and Decision Making

At WWF

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AI-Assisted Decision-Making for Conservation, October 20, 2022

WWF works in 100 countries around the world



Algeria Argentina Armenia Australia Austria Azerbaijan Belgium Belize Bhutan Bolivia Bosnia Herzegovina Brazil Bulgaria Cambodia Cameroon Canada Central African Republic Chile China Colombia Comoros Croatia Cuba Democratic Republic of Congo Denmark Ecuador Fiji Finland France French Guyana Gabon Georgia Germany Greece Guatemala Guyana Honduras Hong Kong Hungary India Indonesia Italy Japan Kazakhstan Kenya Korea Kyrgyzstan Laos Latvia Madagascar Malaysia Mexico Mongolia Morocco Mozambique Myanmar Namibia Nepal Netherlands New Caledonia New Zealand Norway Pakistan Panama Papua New Guinea Paraguay Peru Philippines Poland Portugal Republic of Congo Romania Russia Serbia Seychelles Singapore Slovakia Solomon Islands South Africa Spain Suriname Sweden Switzerland Tajikistan Tanzania Thailand Tunisia Turkey Turkmenistan Uganda United Kingdom Ukraine United Arab Emirates United States Uzbekistan Vietnam Zambia Zimbabwe

Where we work



Wildlife



Forests



Oceans



Freshwater



Climate & Energy



Food



Finance



Markets



Governance

THE NEW DEAL FOR NATURE AND PEOPLE

STOP AND REVERSE THE DOWNWARDS DECLINE OF BIODIVERSITY TO PROTECT AND RESTORE NATURE BY 2030, FOR THE BENEFIT OF PEOPLE AND THE PLANET, IN SUPPORT OF THE SUSTAINABLE DEVELOPMENT GOALS.

THE PROBLEMS

MASSIVE
DESTRUCTION OF
NATURAL SPACES



MASS LOSS
OF SPECIES AND
DIVERSITY OF LIFE



UNSUSTAINABLE
PRODUCTION
AND CONSUMPTION
OF NATURE'S RESOURCES



THE TARGETS

ZERO
LOSS
OF NATURAL
HABITATS



ZERO
HUMAN-
INDUCED
EXTINCTION



HALVE
FOOTPRINT OF
PRODUCTION
AND
CONSUMPTION



THE SOLUTIONS

- PROTECT 30%
- SUSTAINABLY MANAGE AT LEAST 20%
- RECOGNIZE RIGHTS TO INDIGENOUS PEOPLES' LANDS



- STOP UNSUSTAINABLE WILDLIFE EXPLOITATION AND TRADE
- ENABLE VIABLE POPULATIONS



- TRANSITION TO SUSTAINABLE PRACTICES: INFRASTRUCTURE, AGRICULTURE, FISHING, EXTRACTIVES

THE BENEFITS



WATER FOR
9 BILLION



FOOD FOR
9 BILLION



DIVERSITY
OF LIFE



STABLE
CLIMATE



GOOD QUALITY
OF LIFE



Many actors: Public, Private, NGO, Individuals, Funders

Many scales: Spatial, temporal

Many indicators: 1000s, at least

Many types of decisions - internal and external

CMP Conservation Actions Classification v 2.0

A. TARGET RESTORATION / STRESS REDUCTION ACTIONS

1. Land / Water Management

- 1.1 Site/Area Stewardship
- 1.2 Ecosystem & Natural Process (Re)Creation

2. Species Management

- 2.1 Species Stewardship
- 2.2 Species Re-Introduction & Translocation
- 2.3 *Ex-Situ* Conservation

B. BEHAVIORAL CHANGE / THREAT REDUCTION ACTIONS

3. Awareness Raising

- 3.1 Outreach & Communications
- 3.2 Protests & Civil Disobedience

4. Law Enforcement & Prosecution

- 4.1 Detection & Arrest
- 4.2 Criminal Prosecution & Conviction
- 4.3 Non-Criminal Legal Action

5. Livelihood, Economic & Moral Incentives

- 5.1 Linked Enterprises & Alternative Livelihoods
- 5.2 Better Products & Management Practices
- 5.3 Market-Based Incentives
- 5.4 Direct Economic Incentives
- 5.5 Non-Monetary Values

C. ENABLING CONDITION ACTIONS

6. Conservation Designation & Planning

- 6.1 Protected Area Designation &/or Acquisition
- 6.2 Easements & Resource Rights
- 6.3 Land/Water Use Zoning & Designation
- 6.4 Conservation Planning
- 6.5 Site Infrastructure

7. Legal & Policy Frameworks

- 7.1 Laws, Regulations & Codes
- 7.2 Policies & Guidelines

8. Research & Monitoring

- 8.1 Basic Research & Status Monitoring
- 8.2 Evaluation, Effectiveness Measures & Learning

9. Education & Training

- 9.1 Formal Education
- 9.2 Training & Individual Capacity Development

10. Institutional Development

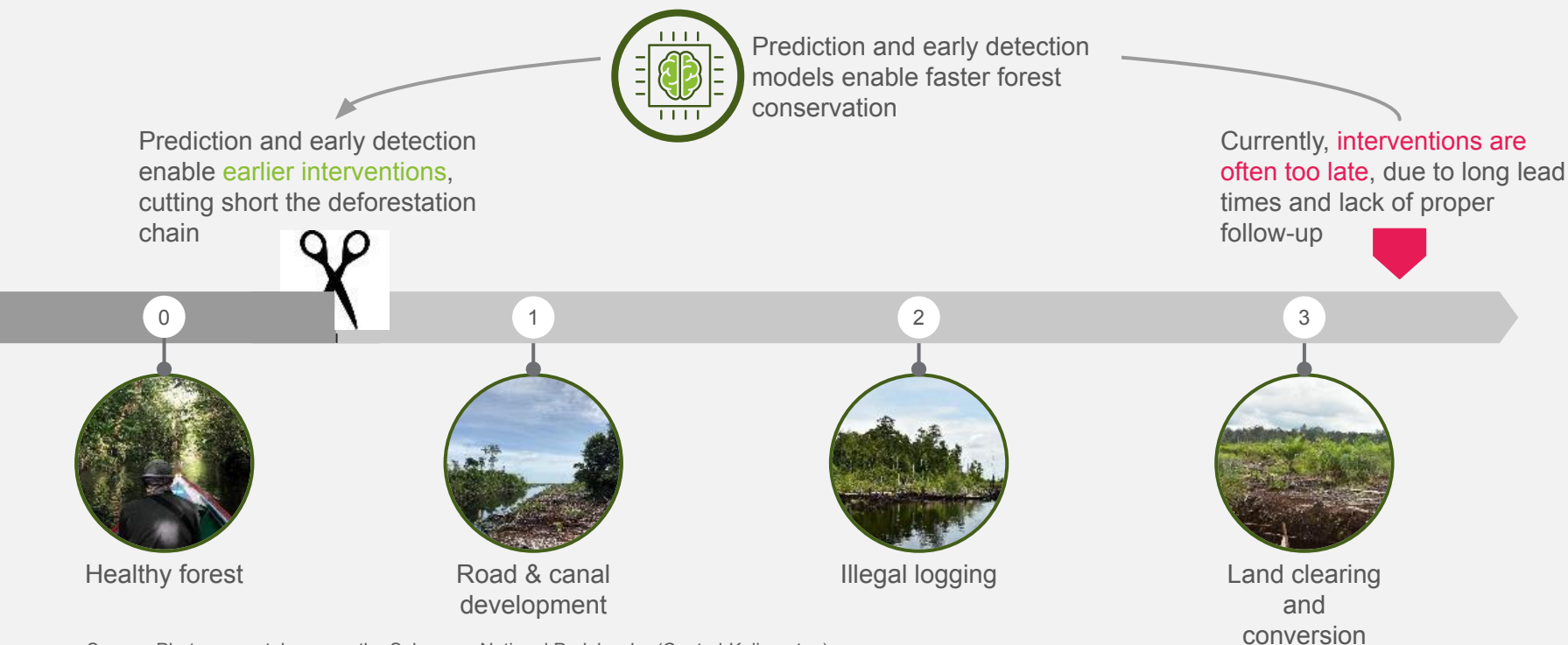
- 10.1 Internal Organizational Management & Administration
- 10.2 External Organizational Development & Support
- 10.3 Alliance & Partnership Development
- 10.4 Financing Conservation



Forest Foresight

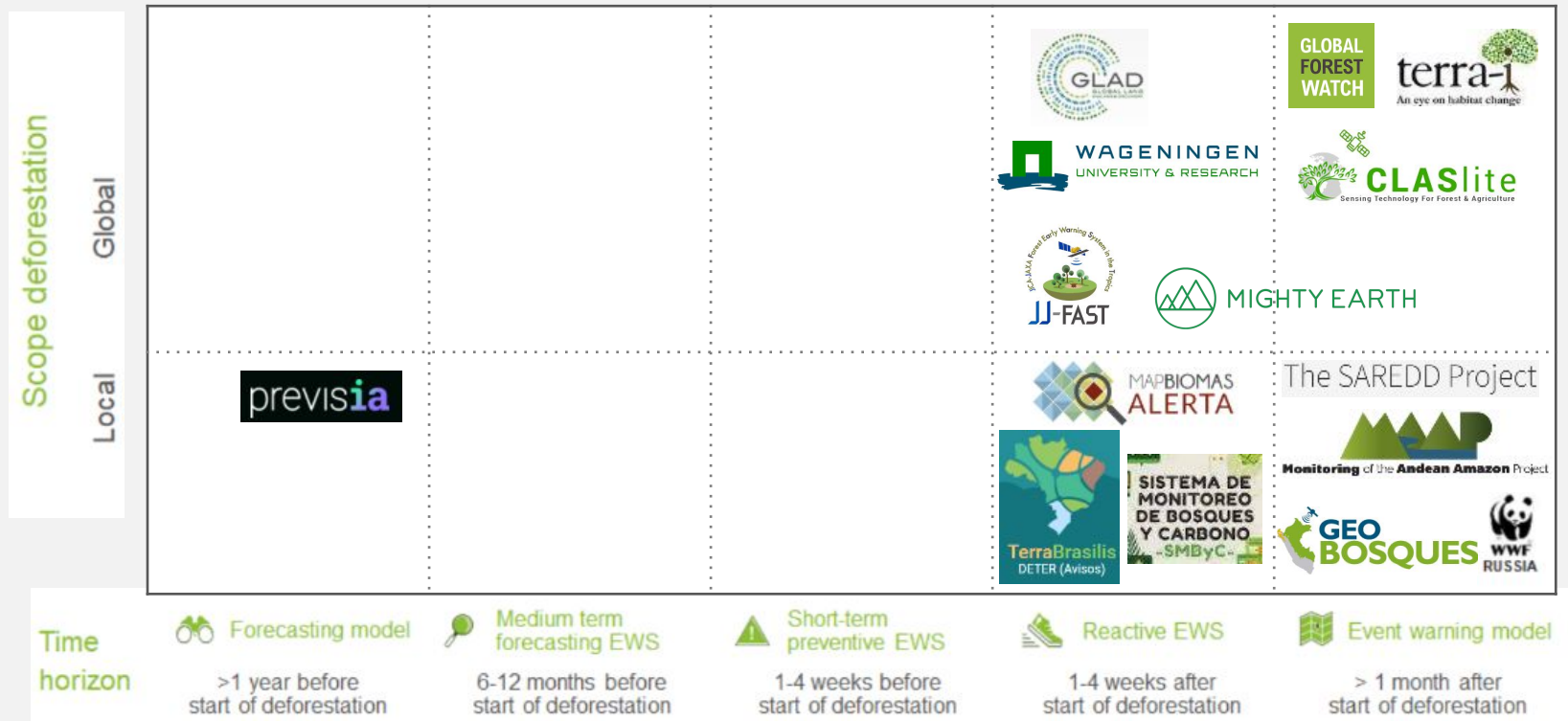
An early warning system to
predict and prevent
deforestation

The power of prediction: enabling earlier interventions for forest conservation

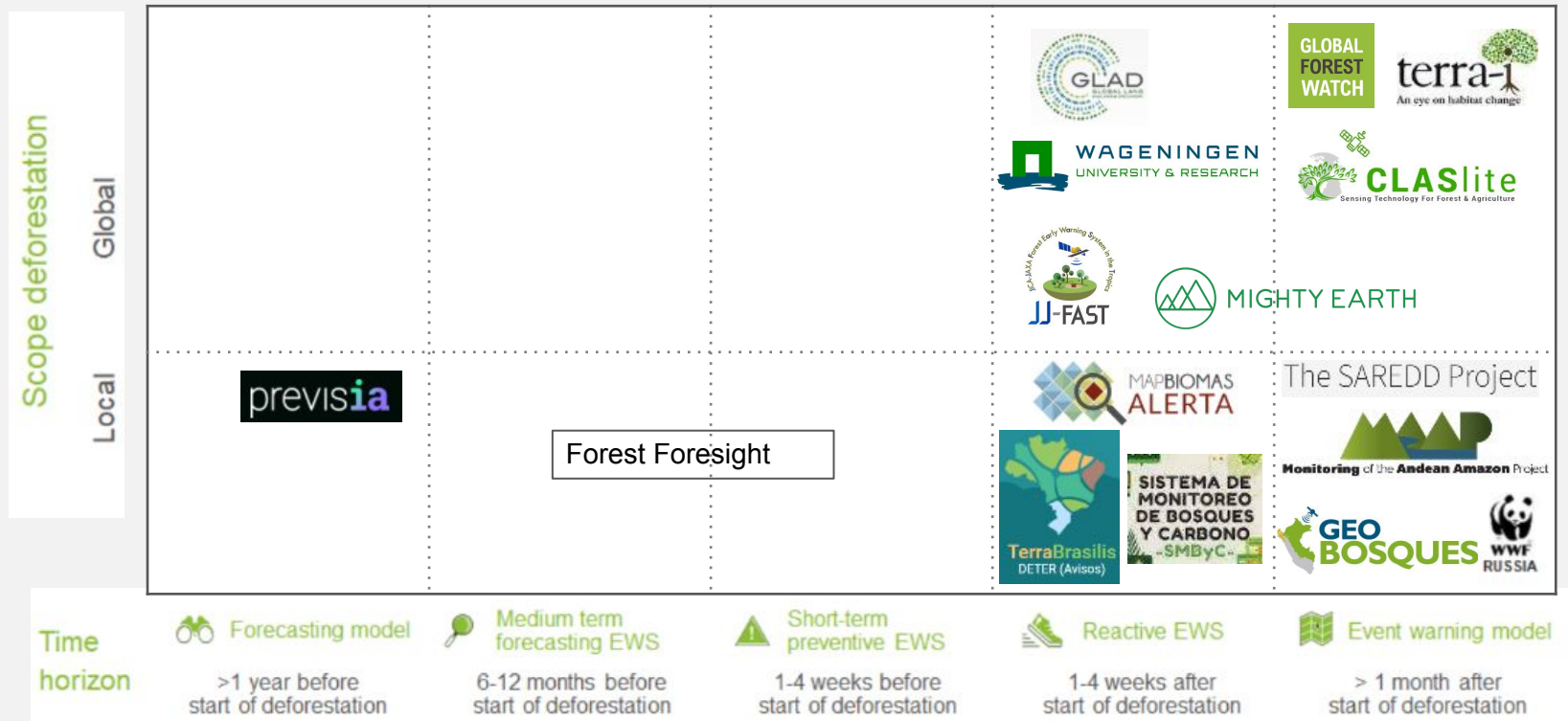


Source: Photos were taken near the Sebangau National Park border (Central-Kalimantan)

Existing forest monitoring systems



Existing forest monitoring systems



Forest Foresight: from data to interventions and tools to support rollout

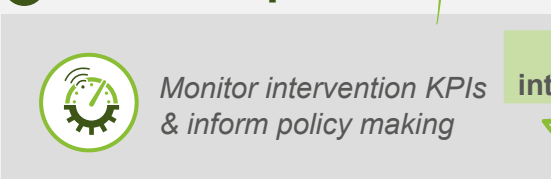
1 Generate prioritized deforestation alerts



2 Follow up alerts



3 Measure impact

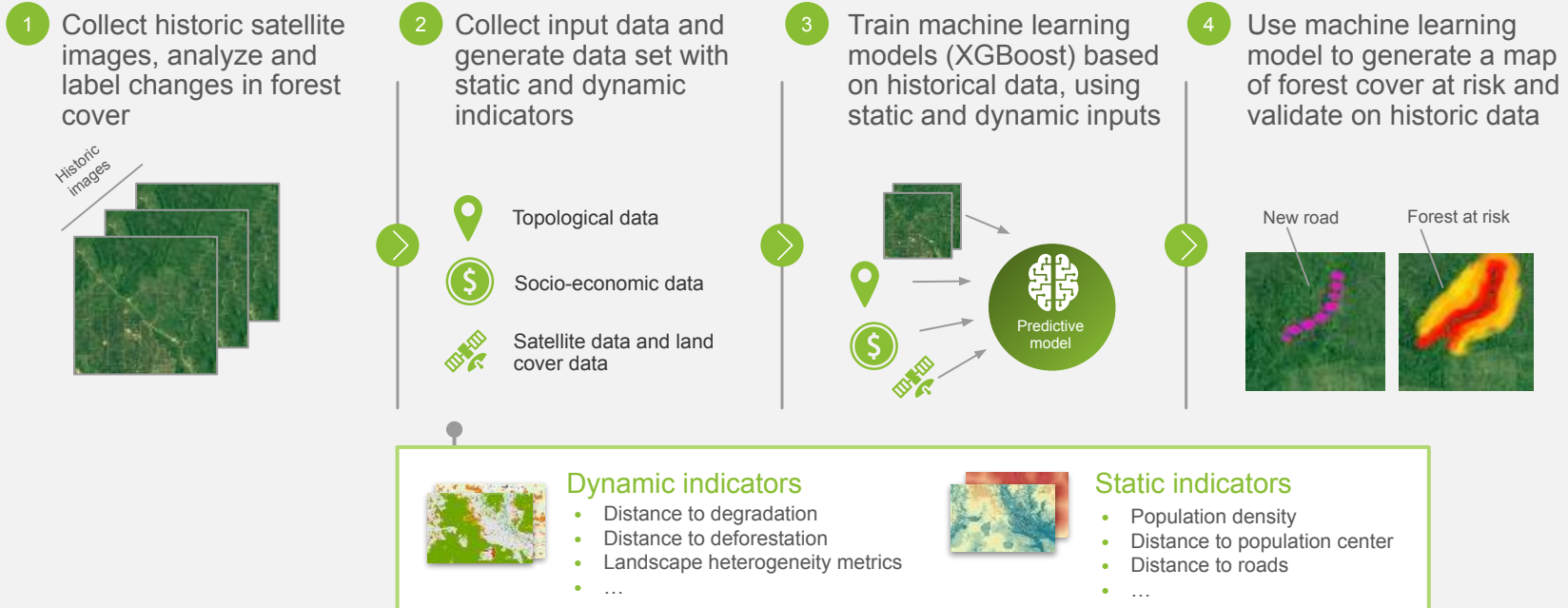


1. Predictions on 15x15m, followed by down-sampling to 480x480m, 2. Exclude deforestation with permits, forest fires etc., 3. Includes desktop research to verify permits, area mgmt. plan etc.









Technical foundation underlying FF



*Predict hot
zones at risk*



Several open-source indicator generators used, local data added after

Indicator generator		For dynamic or static data?	Definition	Output indicator	Source
Distance		Either	Calculates Euclidean distance from a vector or raster feature	Distance from target feature (e.g. roads, waterways)	SciPy
Slope of terrain		Static	Computes slope of land from digital elevation model	Slope	NumPy
Indicator formatters (x 9)		Either	Series of tools that process indicators into usable format for data pipeline	Many (e.g. elevation, population density, month of deforestation)	OpenCV, SciPy, NumPy
Percent land cover ¹		Dynamic	Proportion of total landscape area occupied by each land cover class	For each land cover class, % of class in 1km ² grid cell	McGarigal 2015
Edge density ¹		Dynamic	Proportion of the lengths of all edge segments in the landscape to the total landscape area	For each land cover class, meters per hectare of patch edge in 1km ² grid cell	McGarigal 2015
Patch density ¹		Dynamic	Number of patches of each land cover patch type divided by total landscape area	For each land cover class, number of patches per 100 hectares in 1km ² grid cell	McGarigal 2015
Shannon's Diversity Index ¹		Dynamic	Proportion of the landscape occupied by each land cover patch type	For all classes, Shannon's Diversity Index in 1km ² grid cell	Shannon and Weaver 1949
Aggregation Index ¹		Dynamic	Proportion of maximum like adjacencies between pixels of each land cover patch type based on the single-count method.	For each land cover class, percent of like, single-count adjacencies in 1km ² grid cell	He et al. 2000

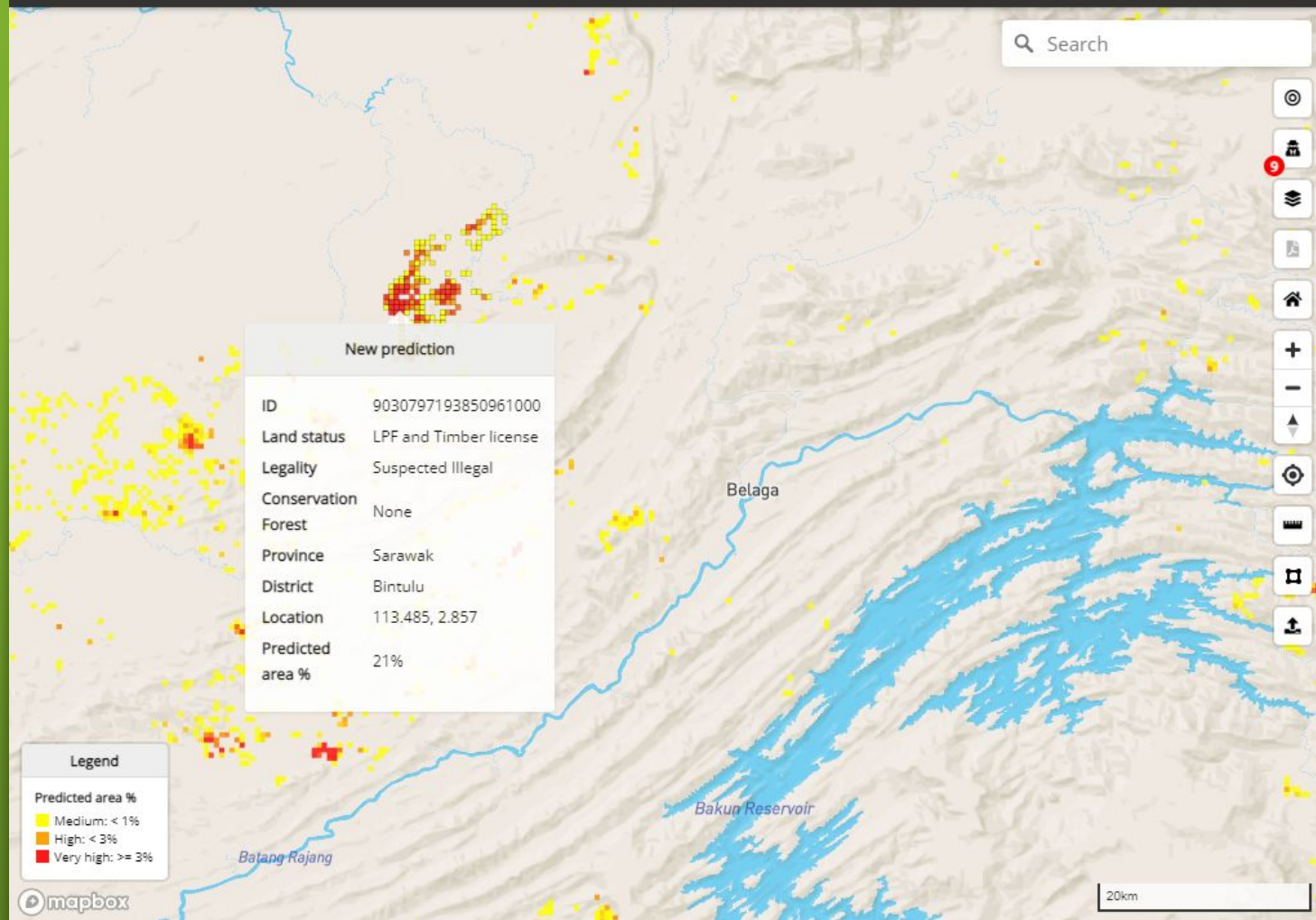
1. Definitions and open-source python versions adapted from [University of Massachusetts Amherst FRAGSTATS 4.2 statistical pattern analysis program](#)
Important FRAGSTATS indicators selected from Cushman et al. 2017

Additional data sources beyond satellite data

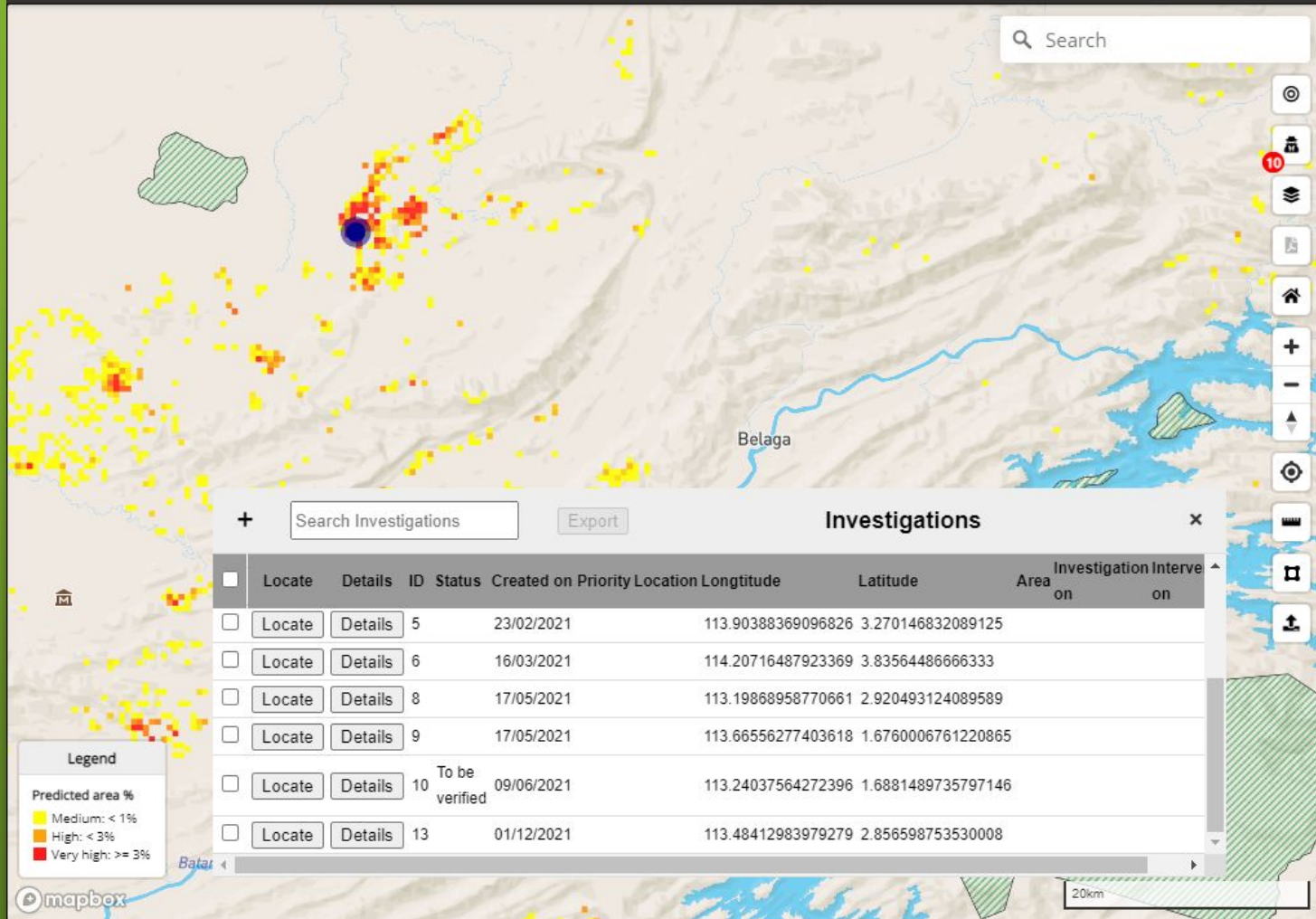
Data Source	Description	Typical example	Purpose
<i>Deforestation change detection</i>	Tracks deforestation every 6-12 days, typically very detailed (e.g. 30mx30m)	University data, purchased data	<ul style="list-style-type: none"> Predict deforestation Monitor deforestation
<i>Vegetation baseline map</i>	Land cover map	Ministry data	<ul style="list-style-type: none"> Predict deforestation
<i>Geographical datasets:</i> <ul style="list-style-type: none"> Coastline Elevation Admin boundaries (e.g. districts) Roads and Waterways Villages/settlements Population density 	Several geographical datasets that can be used to make better predictions, typically publicly available	ASTER, Open Street Map, Ministry data	<ul style="list-style-type: none"> Predict deforestation
<i>Forest fires</i>	Shows forest fires in last 7 days	NASA	<ul style="list-style-type: none"> Show on dashboard
<i>Land status and legality</i>	Spatial planning to show where deforestation is illegal	Ministry data	<ul style="list-style-type: none"> Show on dashboard Prioritize predictions
<i>Concessions</i> <ul style="list-style-type: none"> Timber Palm oil Cacao Mining 	Concession data to help make better predictions (e.g. near palm oil mills), or show where deforestation is legal	CIFOR, GFW, Ministry data	<ul style="list-style-type: none"> Predict deforestation Show on dashboard
<i>Logging roads</i>	Location of logging roads	CIFOR, Ministry data	<ul style="list-style-type: none"> Predict deforestation
<i>Soil map (e.g. peat)</i>	Soil type maps	Ministry data	<ul style="list-style-type: none"> Predict deforestation
<i>Protected areas, HCVA</i>	Location of protected areas or High Conservation Value areas	WDPA, Ministry data	<ul style="list-style-type: none"> Show on dashboard Prioritize predictions
<i>Species habitats</i>	Species maps	Ministry data or WWF data	<ul style="list-style-type: none"> Prioritize predictions
<i>Carbon stock</i>	Above ground carbon stock	Ministry data	<ul style="list-style-type: none"> Prioritize predictions

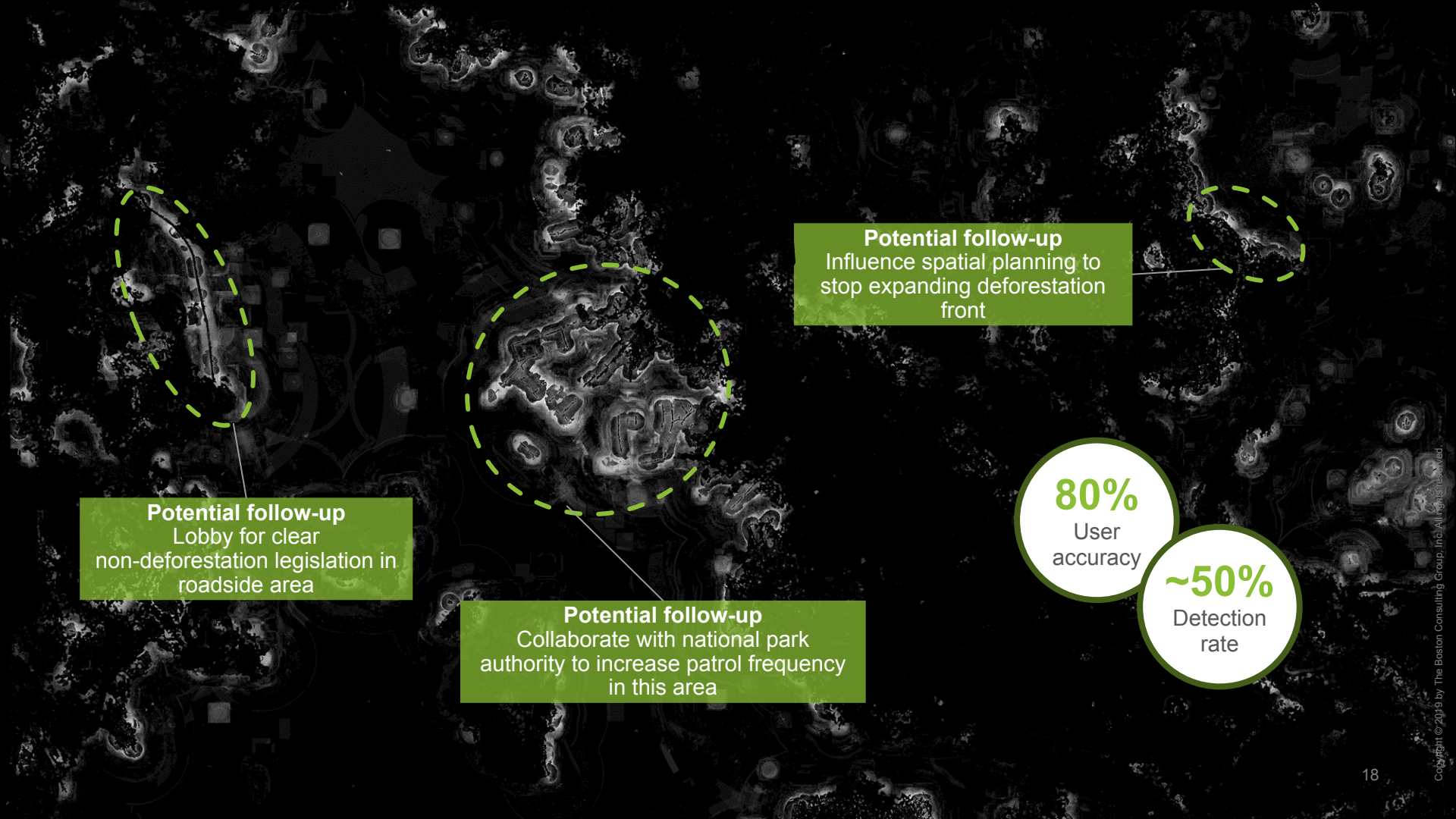
Dashboard visualizes predictions of 'deforestation hot zones'

EARLY WARNING SYSTEM



When potentially illegal deforestation is spotted, users can create investigations to follow up the alert





Potential follow-up
Lobby for clear
non-deforestation legislation in
roadside area

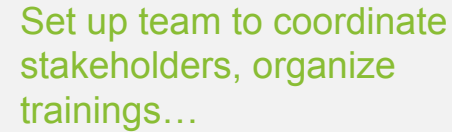
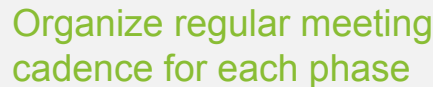
Potential follow-up
Collaborate with national park
authority to increase patrol frequency
in this area

Potential follow-up
Influence spatial planning to
stop expanding deforestation
front

80%
User
accuracy

~50%
Detection
rate

Setup of local governance critical to FF success



FF global team



+ Future partnerships

FF local (Kalimantan)





MOTION DETECTING CAMERAS AND AI



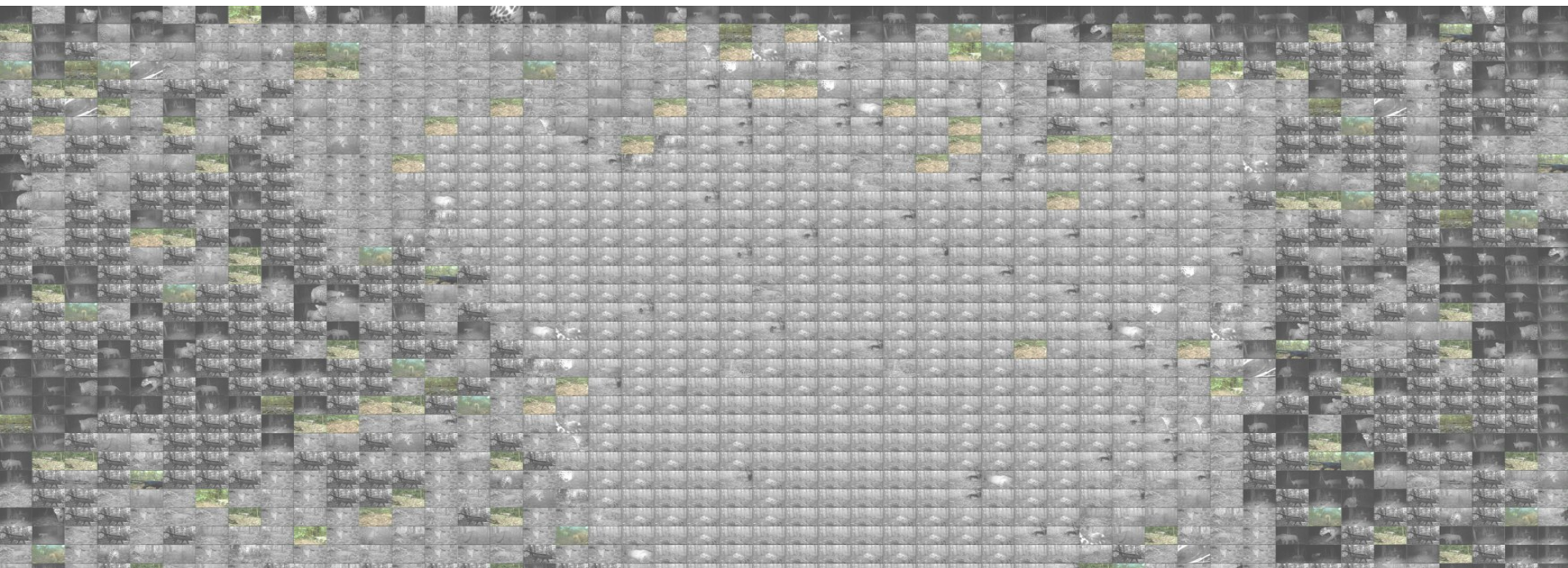
Motion sensitive cameras are everywhere

1,000s of projects

100,000s of cameras

Many 10,000,000s of images

Steenweg et al. 2017



A photograph of a snow leopard and its cub in a mountainous landscape. The adult leopard is in the foreground, looking towards the camera, while the cub is slightly behind and to the left. The background shows a steep, rocky mountain slope under a clear sky.

wildlifeinsights.org

Harness the power of AI for wildlife



Goal

Provide wildlife professionals a software platform to identify, manage and analyze camera trap data



Deploy Field Sensors



Organizations deploy sensors (mostly camera traps) in situ

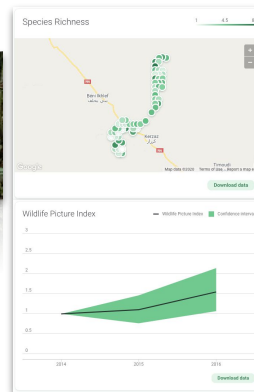
Upload



Images and metadata are uploaded to the Google Cloud, and AI identifies what's in the image. Analytics are automatically calculated.

Identify

Analyze



Manage & Act



Reports are produced to provide insights to decision-makers and stakeholders

Showing **43,070,623 camera trap records** taken in the whole world between 1990-01-02 and 2022-08-07.

[See filters and statistics](#)

Exploring
data



wildlifeinsights.org

Eyes on Recovery

Using Wildlife Insights to measure the impact of the 2019-20 19 million hectare bushfires on Australia's wildlife

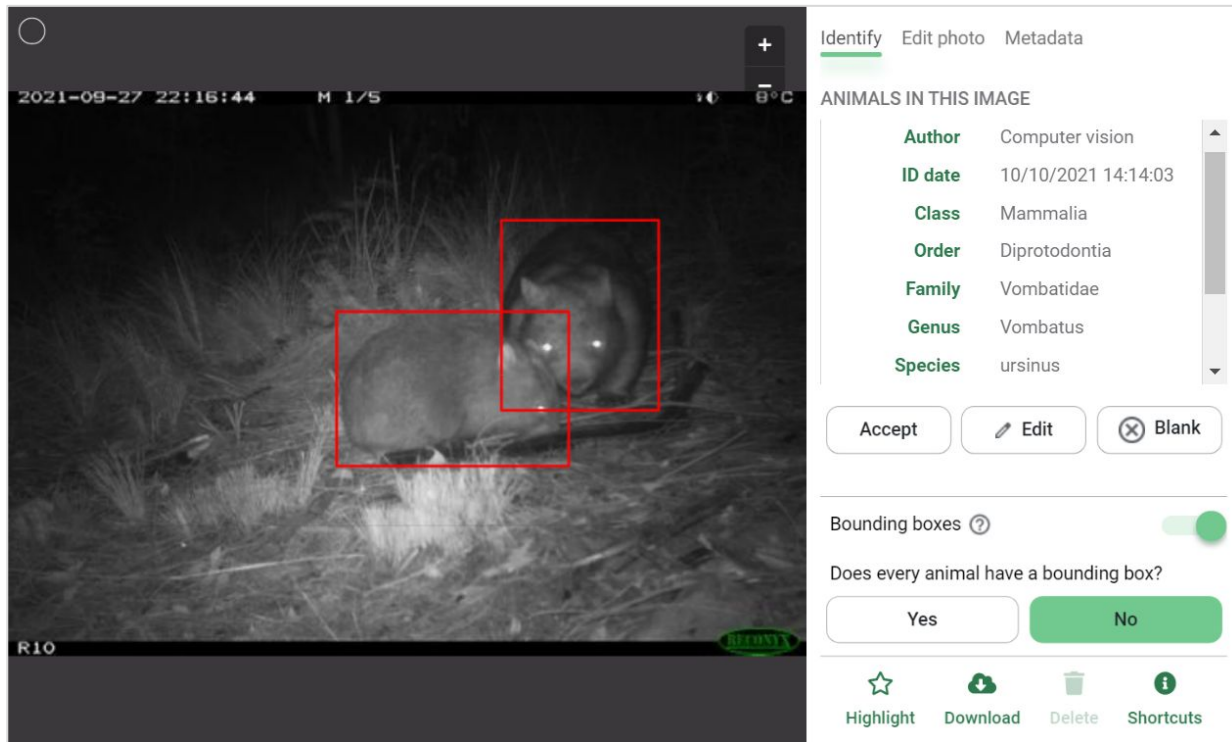


© Kangaroo Island Land for Wildlife



Abby Hehmeyer & Arno Lyet (WWF US), Emma Spencer (WWF AUS)

Data processed in Wildlife Insights



2.5 million images used to train WI computer vision model for Australia

Excellent feedback from AUS users: **5x increase in image processing speed**

~500 hours of staff time saved per project site location



Be part of the biggest change in the Colombian Amazon

Connecting investors, donors and philanthropists with carefully
identified investment opportunities.



Search projects, investors, open calls...

Filters

Search



ARIES

Artificial Intelligence for Environment & Sustainability

bc³

BASQUE CENTRE
FOR CLIMATE CHANGE
Klima Aldaketa Ikergai

Sustainability, that's it!

Interoperability: A core challenge to the global ecosystem services community



FINDABLE

Data has rich metadata and unique identifier



ACCESSIBLE

Data can be easily downloaded or used by using standard protocols



INTEROPERABLE

Metadata use an accessible and standard language



REUSABLE

Data is well-described and provides clear usage of licences



1. Semantics

The language used to describe scientific observations must be flexible and shareable, without ambiguity. It must efficiently address all the “W’s of information – *what, where, when, why, and how* – without becoming too large or complex to learn and use.



3. Open, linkable models

While today’s dialogue on interoperability focuses on data, it can help to see models and model components as other ways to make scientific observations. This enables a single, consistent discussion on how to semantically connect data to



2. Open, linkable data

Making data and models [FAIR](#) is complex and requires understanding of – and agreement on – the nature of *all* scientific information, including, but not limited to, adopting URIs and open standards.



4. Software infrastructure

We believe in contributions that can start making positive change immediately. Our theoretical semantic integration work is complemented by an [open-source software stack](#) built and maintained by the Partnership. The software



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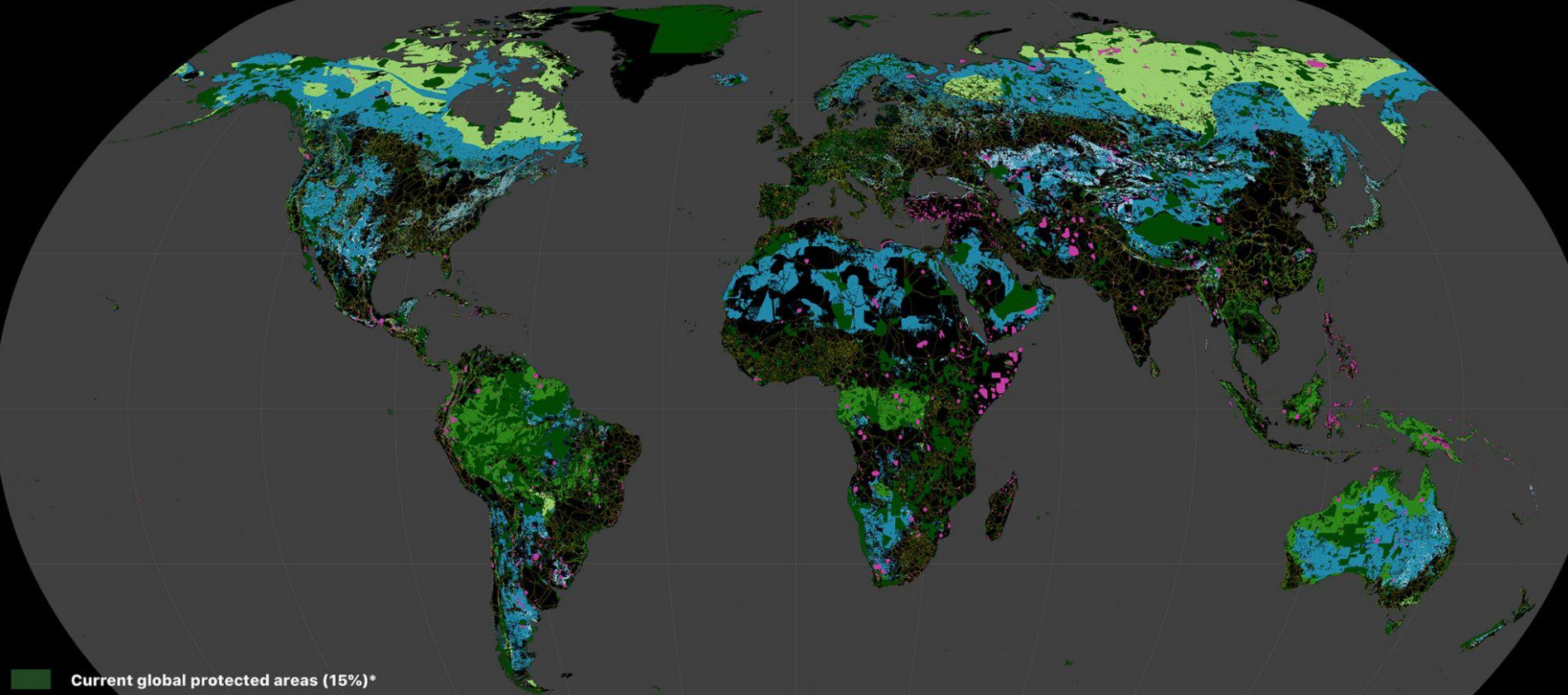
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Comments



Map credits © OSM contributors.

Moving From Where to How

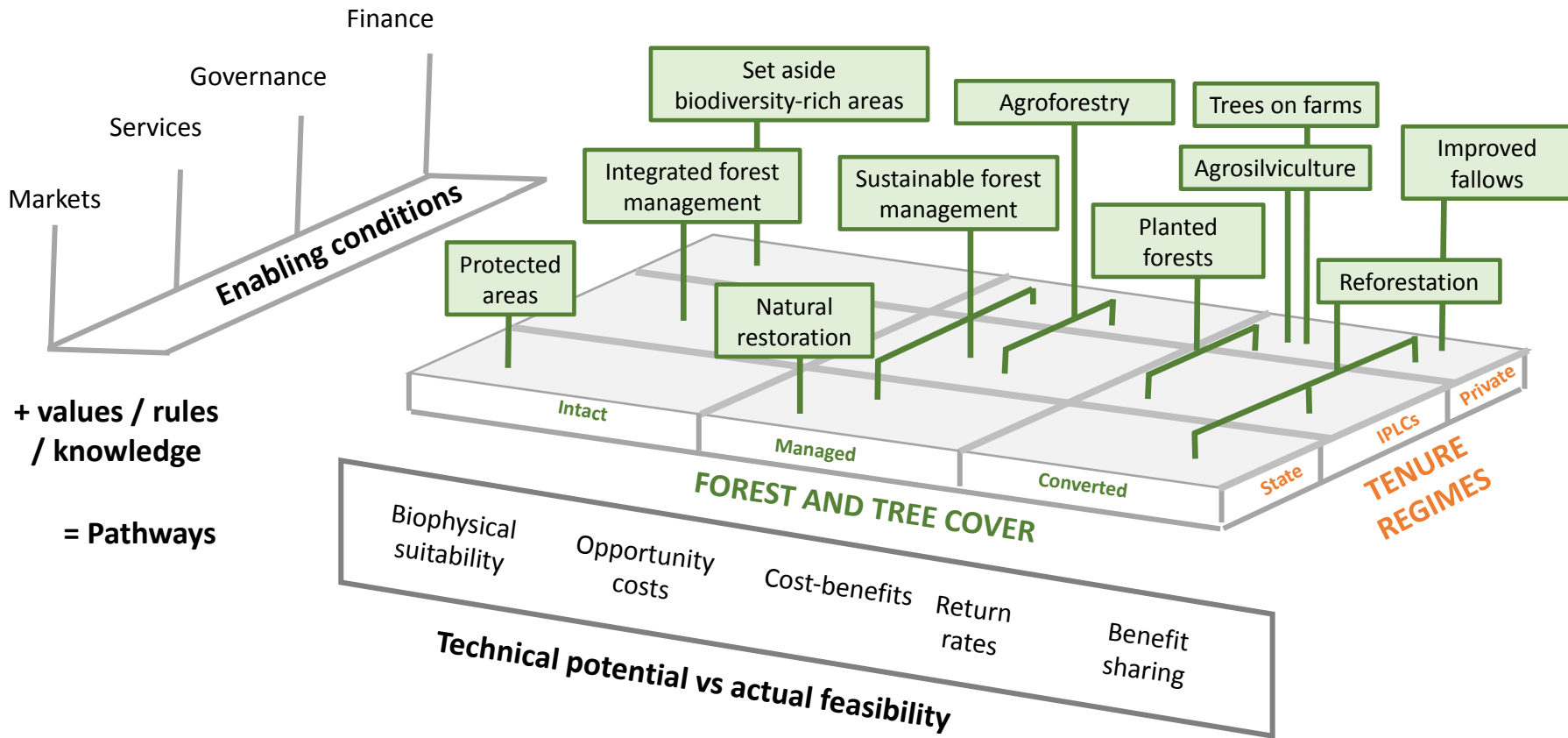


- Current global protected areas (15%)*
- Additional unprotected areas needed to conserve sites of **Species Rarity** (2.3%)
- Additional unprotected areas needed to conserve sites of **Distinct Species Assemblages** (6.0%)
- Additional unprotected areas needed to conserve sites of **Rare Phenomena** (6.3%)
- Additional unprotected areas needed to conserve sites of **Intactness** (16.0%)
- Additional **Climate Stabilization Areas** (4.7%)
- Wildlife and Climate Corridors

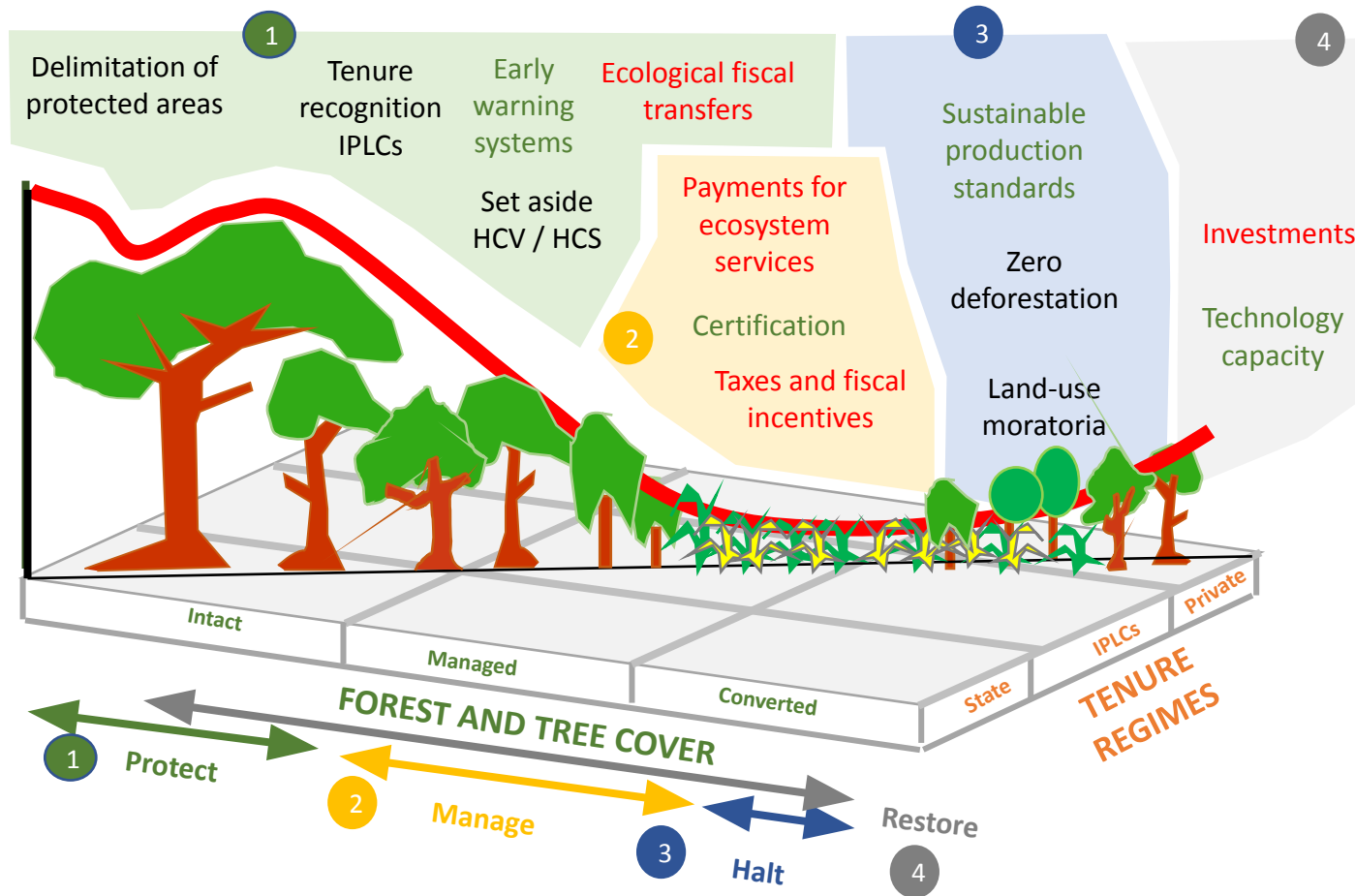
* Including polygons selected for Species Rarity, Distinctness, Rare Phenomena, and Intactness.

<https://www.globalsafetynet.app/>

MULTIPLE FOREST-BASED OPTIONS, POTENTIAL AND FEASIBILITY



MEASURES ADAPTED TO PLACE-SPECIFIC CONDITIONS



Examples of measures



State regulations,
private policies,
and hybrid
initiatives



. Institutional
. Technical
. Economic



DISCOVER □

ACT □

LATEST □

60 YEARS OF ACTION FOR NATURE AND PEOPLE

As part of our 60th anniversary

Big Outstanding Challenges

- Social Metrics (e.g. poverty, climate resilience, language)
- Data privacy and sovereignty versus FAIR principles
- Connecting Global and Local (See GLASSNET for example)
- Digital Sustainability (See the Coalition for Digital Environmental Sustainability - [CODES](#))
- The big one - Impacts: measuring and inferring what works and why



Thanks!



WWF