

WILDLIFE INSIGHTS

A platform to maximize the potential of camera trap
and other passive sensor wildlife data for the planet

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Conservation International, Moore Center for
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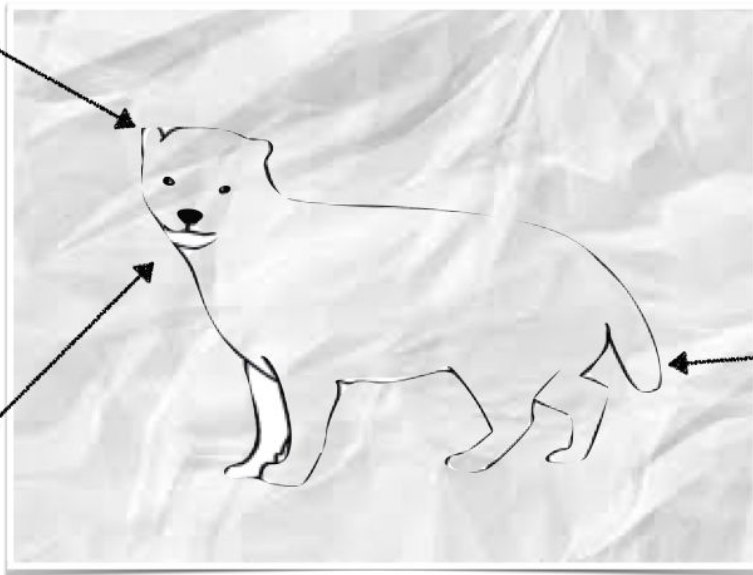
Jonathan Palmer

Wildlife Insights Steering Committee
Wildlife Conservation Society, Executive Director
Conservation Technology

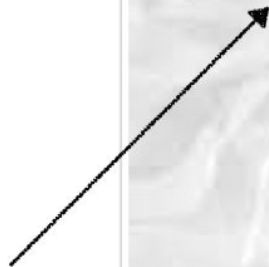




Short ears



Flattened snout



Short tail



2014-09-15 8:26:25 AM M 3/3

20°C



CCT-YAN-2-05

RECONYX

@ TEAM Network

A big challenge in wildlife conservation

Data: reliable, current, verifiable

Derive insights from data quickly

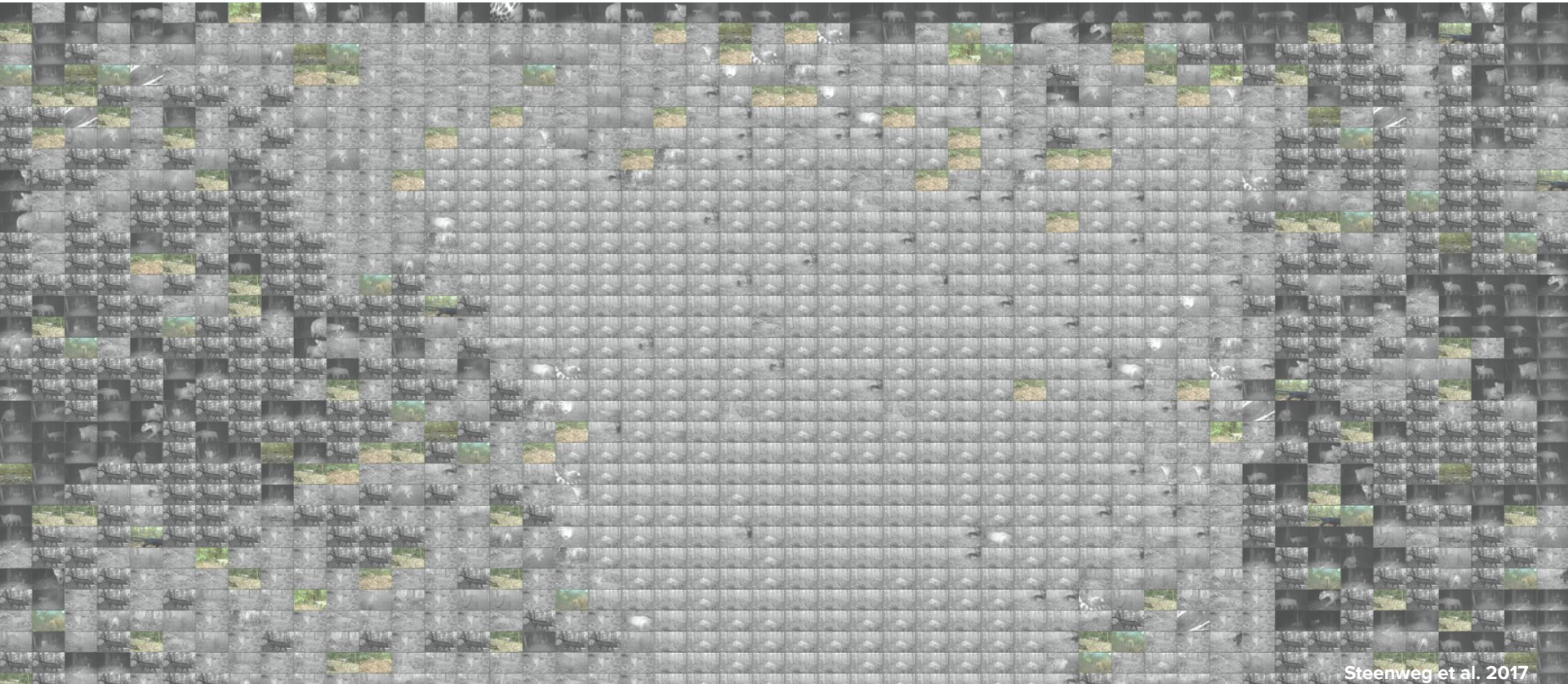
**Collected by multiple organizations
and individuals around the globe**

Camera traps are everywhere

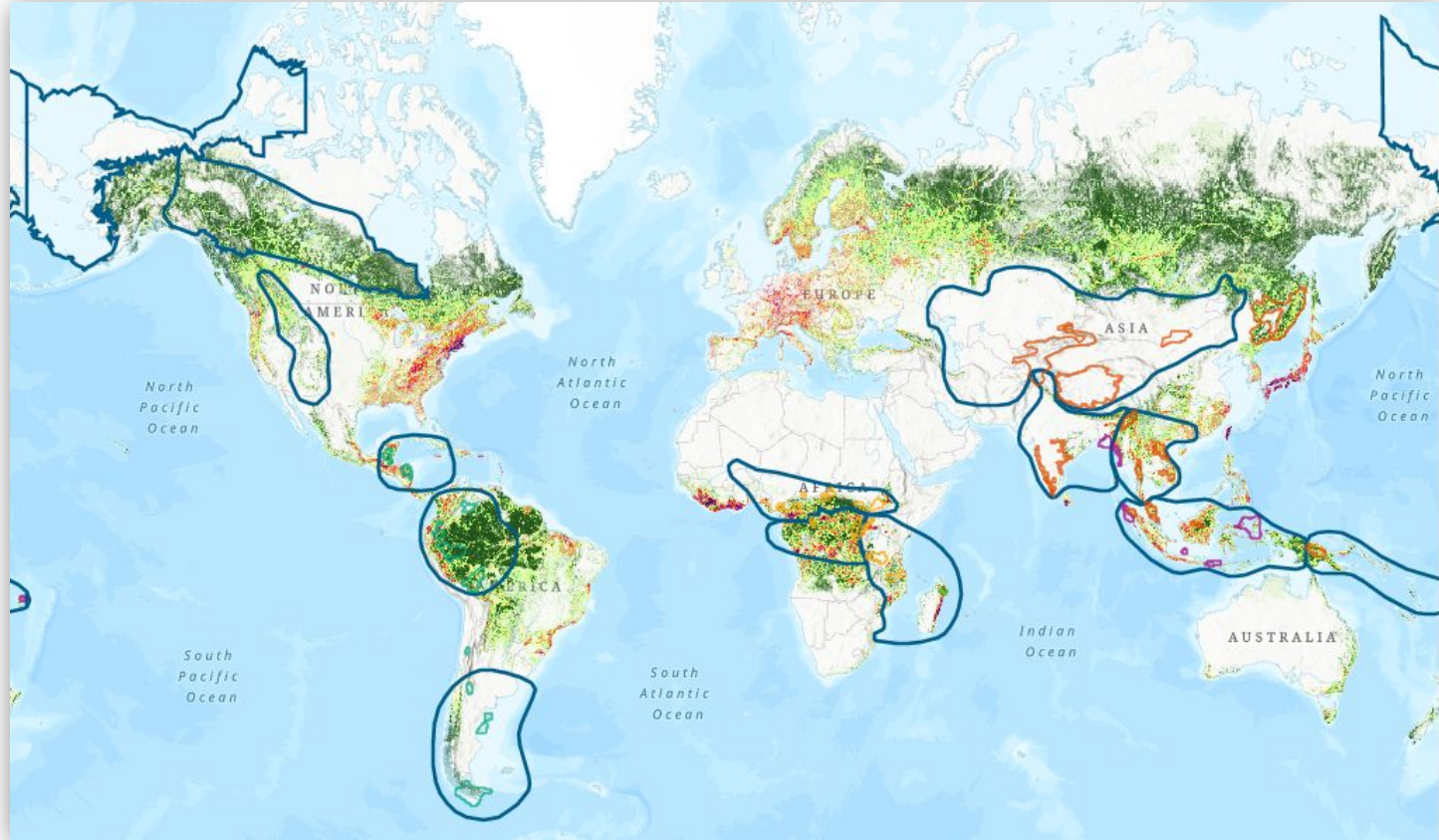
1,000s of projects

100,000s of camera traps

1,000,000s of images



WCS Background

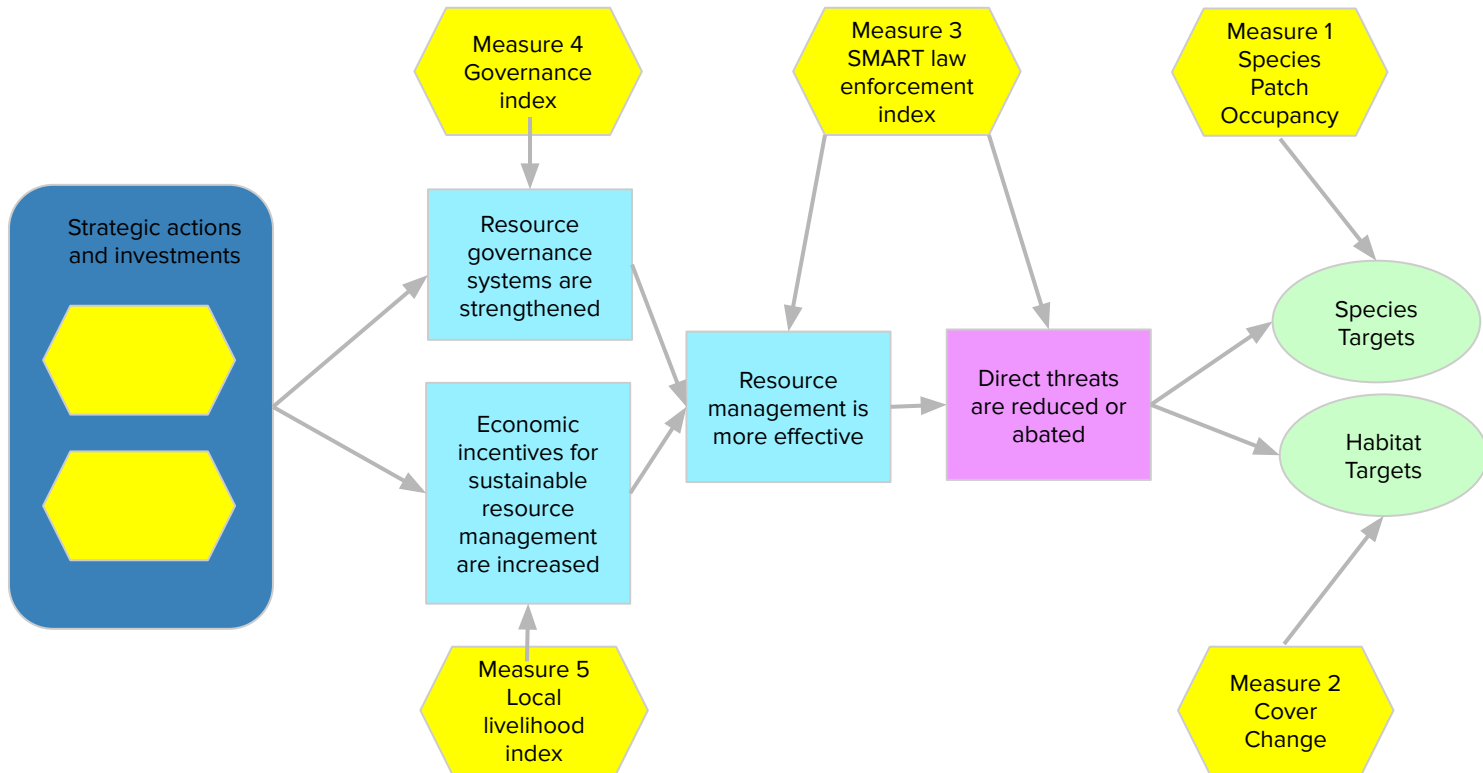




A commitment to Solving Conservation Issues in Partnership and at Scale



Tying Biodiversity Monitoring to Action





WCS Camera Trap Background

3+ million catalogued images

40 - 100 mn images in the wild

≈ 1000 camera traps deployed

No standard camera trap data curation



The Potential Value of Wildlife Insights

Time saved = time doing conservation

Out of the box analytics

Integration & the bigger picture

More current population estimates

Major Barriers

1
Slow data flow

2
Data siloed
and raw

3
No tools to
gain insights

Solutions

1

**Software 10x better,
Artificial Intelligence**

2

Modern platform

3

**Data-driven
insights**



Wildlife Insights



Our Vision and Principles

**We envision a world
where wildlife
populations are stable
or recovering**

Data-driven actions

Integrity

Partnership

Transparency

Open access to data

Key Features



Artificial Intelligence

- 614 species trained on 8.7M images
- Identifies 79% of blank images with <2% error
- 100 common species have 80-98.6% probability of correct prediction
- Including a detector and bounding boxes in training
- Leveraging sequential information



Data input and management

- Web-based uploads and management to review identifications
- Batch uploads for legacy data
- API
- Synced offline desktop solution



Collaboration and Sharing

- Initiatives with personalized websites
- Public page to explore projects
- Creative Commons licenses to share data
- Endangered species safeguarded, no human images shared
- Embargo on projects
- Private data download
- Public data download with DOIs



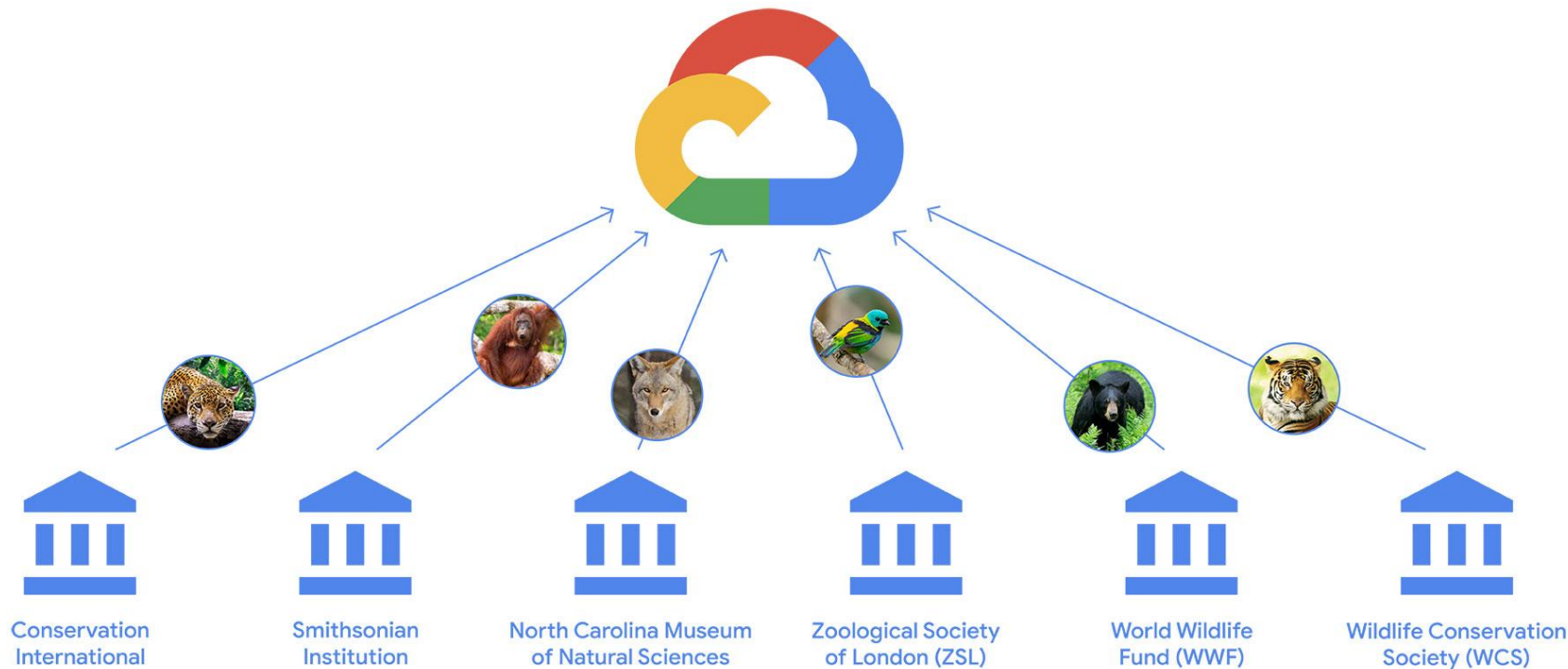
Analytics and Insights

- Operational statistics
- Basic population and activity analytics
- Community wide analytics (e.g., Wildlife Picture Index)
- Automatic report generation
- Spatially explicit products

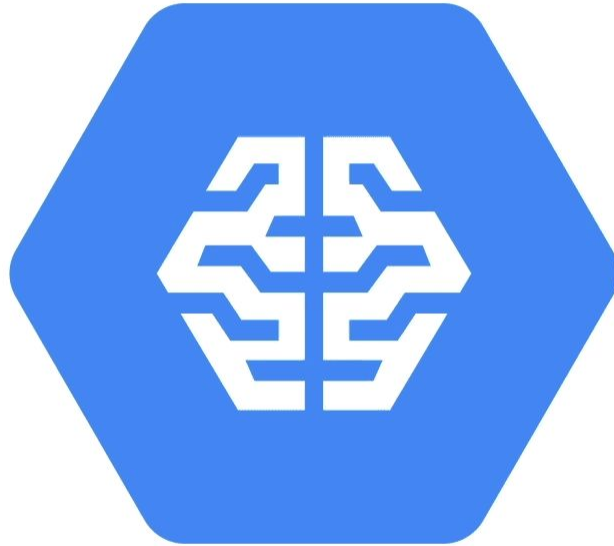
Can AI help identify animals in the wild?



Building good AI models requires lots of labelled data



Training Deep Convolutional Neural Nets

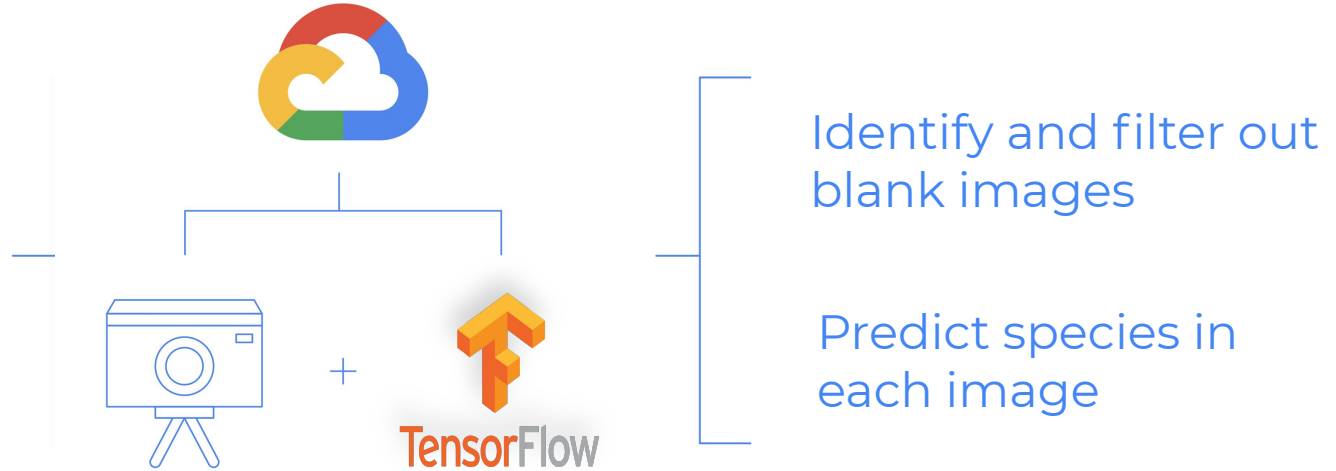
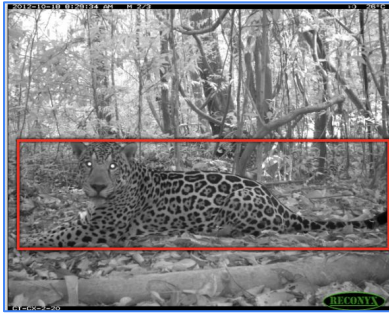



Predicting wildlife species



Wildlife Insights AI model

Predicting wildlife species identifications



A white tiger is standing in a dark, dense forest at night. The tiger is positioned on the left side of the frame, facing right. The background is filled with dark, tangled branches and foliage, creating a mysterious and wild atmosphere. The lighting is low, highlighting the tiger's white fur against the dark surroundings.

The AI model in Wildlife Insights is
trained on **614 wildlife species**
across **8.7M images**.



Camera traps present challenges for AI models



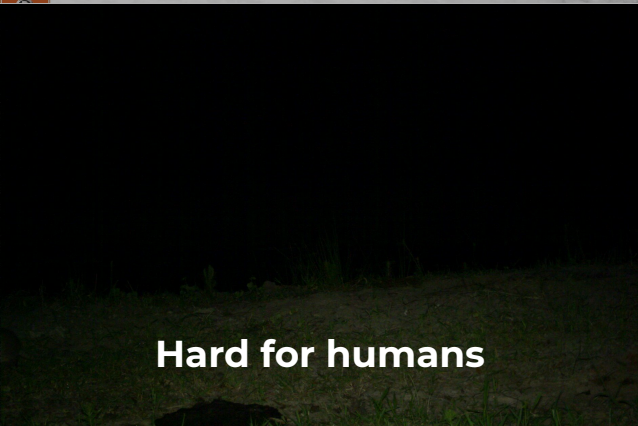
Overexposed



Summer/winter fur
(Showshoe hare)



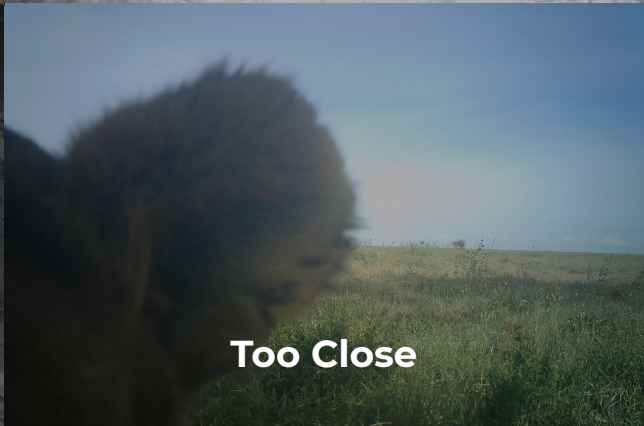
Blurry



Hard for humans



Partial animal



Too Close



Biologist: “I waste so much time
looking at pictures of blowing grass.”

The AI models in Wildlife Insights
catch **79%** of blank images
with an **error rate of less than 2%.**



Biologist: “I don’t want to spend time looking at yet another spotted paca image.”

For ~100 species Wildlife Insights AI models are able to identify between **80%** and **98.6%** correctly.



Human experts can label between
300-1000 images per hour.

A single machine can identify 18,000
images per hour, and when we parallelize
across hundreds or thousands of GPUs,
AI can save biologists a lot of time.

How is the model doing on your species of interest?

<https://www.wildlifeinsights.org/about-wildlife-insights-ai>

Search

Per-Class Performance							
Family	Genus	Species	Common Name	# of images	% of Dataset	Precision	Recall
			blank	3757803	43.07190	98.3	78.7
bovidae	connochaetes	taurinus	common wildebeest	252987	2.89974	92.6	77.4
hominidae	homo	sapiens	human	225966	2.59002	84.8	75.2
cervidae	odocoileus	hemionus	mule deer	211995	2.42989	93.3	80.8
dasyproctidae	dasyprocta	punctata	central american agouti	200068	2.29318	90.8	64.9
suidae	sus	scrofa	wild boar	189713	2.17449	91.2	81.2

Start 1 to 6 of 620 entries

◀ Previous Next ▶



How will the AI models improve?

Help build accuracy on your species
by **joining Wildlife Insights.**

Provide correct species IDs in Wildlife
Insights so the AI models continually
improve.

Manage

Projects list

1

Create

All of your photos are part of a project, which in turn is part of an organization. Get started by creating them.

Create...

2

Upload

By uploading your photos, you'll get suggestions from the Computer Vision and you'll get help from your collaborators.

Upload photos

3

Identify

999+

Accept suggestions or manually tag the animals in the photos. You can search by family, genus or species.

Identify images

4

Analyze

Get insights about where species are located, when are they seen and how much this changes over time.

Analyze

Your projects

Search a project



A Wildlife Organization ⓘ



999+



58



8

[Manage](#)

New project

Organization *:

A Wildlife Organization

**Initiative:**

Select an initiative



Leave empty if the field is not applicable.

Project name *:

Jaguar Monitoring Project



25/47 characters.

Website:

https://jaguarmonitoring.org

28/255 characters.

If the project has a dedicated website, please list it here. E.g., https://wildlifeinsights.org

Short name:

40 characters maximum.

A short name that uniquely identifies the project dataset and relates to project code.

Abbreviation:

255 characters maximum.

If your project has an abbreviated name, please list it here

Country *:

Brazil

**Start date *:****End date:**

[Manage](#) / A Wildlife Organization

Jaguar Monitoring Project

Summary

Details

Identify

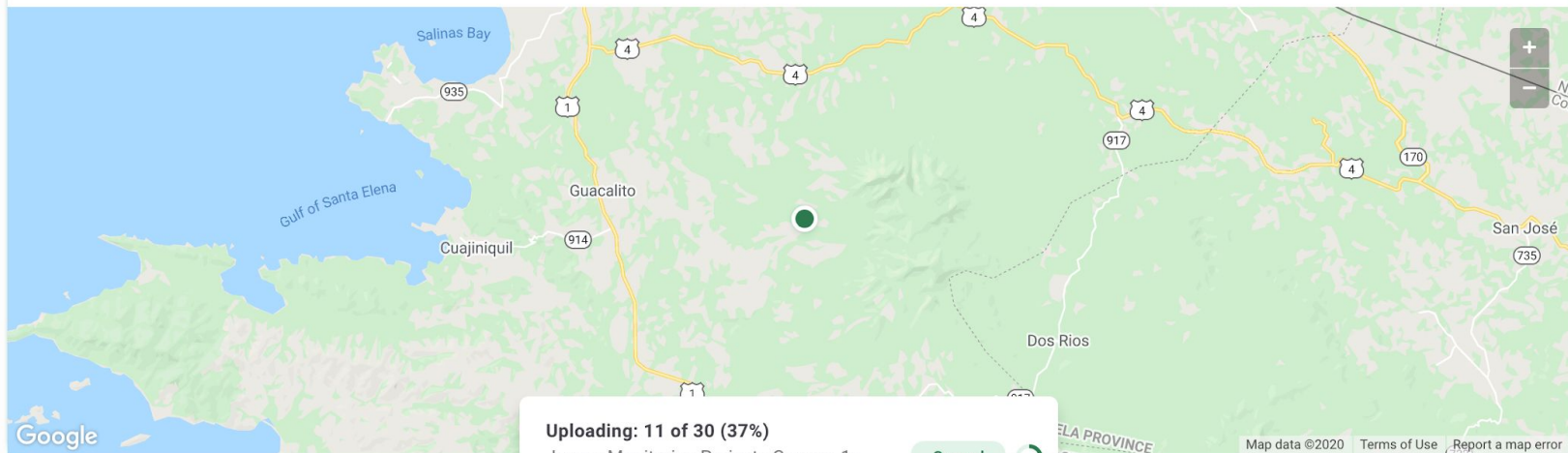
58

Catalogued

Download



Locations



Uploading: 11 of 30 (37%)

Jaguar Monitoring Project - Camera 1a

03/06/2020

Cancel



Map data ©2020

[Terms of Use](#)

[Report a map error](#)

[Manage](#) / [A Wildlife Organization](#)

Jaguar Monitoring Project

Summary Details Identify ⁵³ Catalogued

Download



Camera deployments ▼

Species ▼

Status: All ▼

Photos: All ▼

Date taken ⚙

Bursts: 0 sec





28°C

[Identify](#) [Edit photo](#) [Metadata](#)

Author	Computer vision
Identification date	04/02/2020 08:21
Common name	Giant Anteater
Confidence	98%

[Accept suggestion](#)[Mark as blank](#)[Edit identification](#)[Highlight](#)[Download](#)[Delete](#)

CT-CS-1-28

RECONYX

[Identify](#) [Edit photo](#) [Metadata](#)

Identify this burst of photos by selecting one of the Computer vision's suggestions or add your own identification.

Common name Red Brocket

Confidence 99%

Present in 3 photos

[Mark as blank](#)[Edit identification](#)[Highlight](#)[Download](#)[Delete](#)

CT-CX-1-09





Identify Edit photo Metadata

Brightness



Contrast



Saturation



Presets



Original



High contrast



Brighter



Darker



Highlight

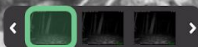


Download



Delete

Projects list

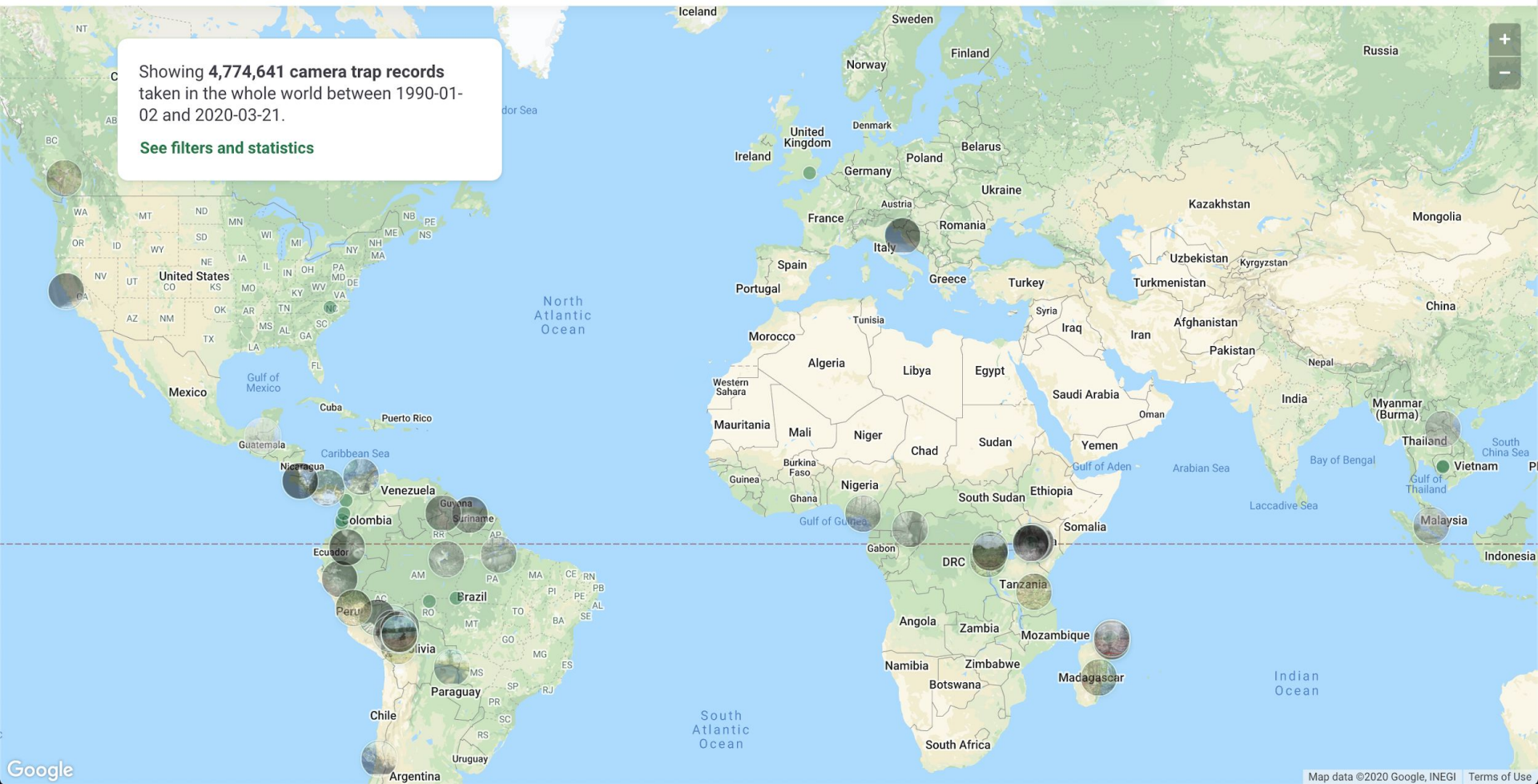


CT-CX-1-09



Showing **4,774,641** camera trap records
taken in the whole world between 1990-01-02 and 2020-03-21.

[See filters and statistics](#)



[Download all data](#)

Showing **4,774,641** camera trap records

of **any species** ▾

in the **whole world** ▾

taken between **1990-01-02** ▾

and **2020-03-21** ▾

and part of **any project** ▾

848

Species



4,774,641

Total images



3,005

Cameras



8,245

Camera deployments





All camera trap records

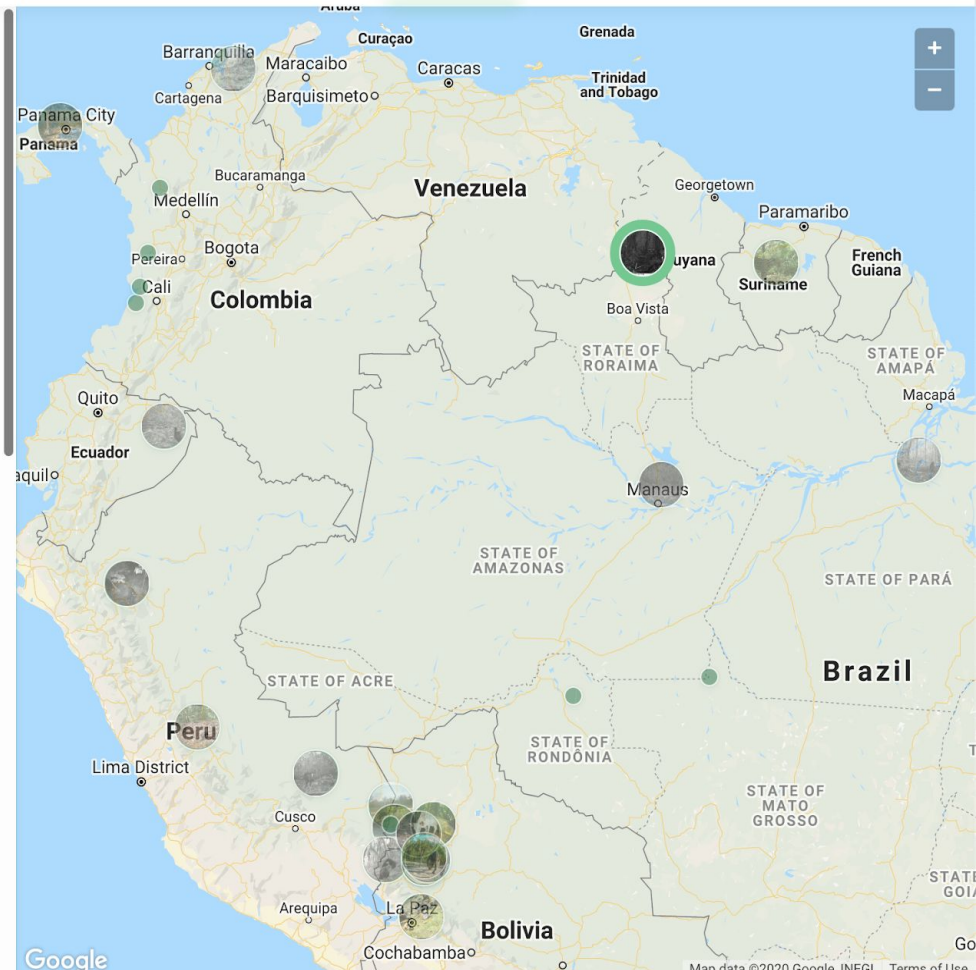
Uei-tepui

ICMBio/CENAP



38

2,404



38

Species



2,494

Total images



1

Cameras



24

Camera deployments



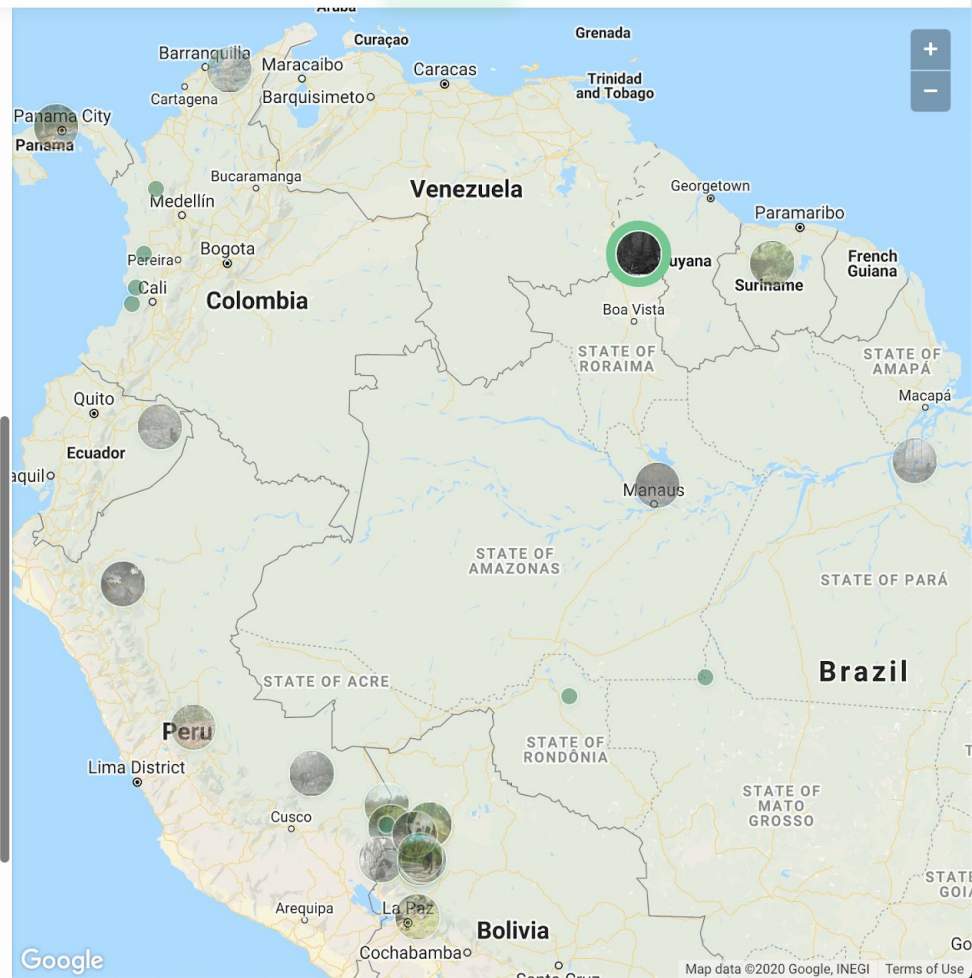
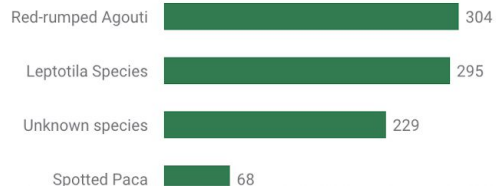
1,417

Wildlife images



Identified species

Count of images per species



But software is not enough



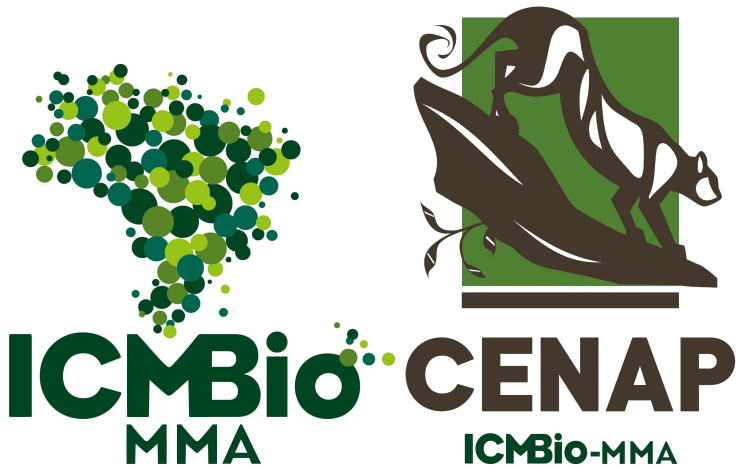
Case Studies



Instituto Chico Mendes de Conservação
da Biodiversidade (ICMBio)
Brazil



Instituto Alexander von Humboldt
Colombia



Instituto Chico Mendes de Conservação
da Biodiversidade (ICMBio)

Brazil

- Central role in providing/advising biodiversity knowledge in the country
- Contributing to national biodiversity reports to the CBD
- Ongoing monitoring of PAs within ARPA
- Adopted camera trap protocol from TEAM in 2016
- Started contributing data to WI last year

ICMBio Dashboard Demo

[LINK](#)

- In charge of managing and streamlining biodiversity data for the national government
- Collecting camera trap data across several sites in the country
- Involved in citizen science efforts and organizing the first national "camera trapping month" across the whole country
- Started contributing data to WI last year



Instituto Alexander von Humboldt
Colombia

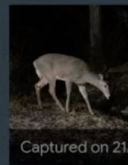
Instituto Humboldt Dashboard Demo

[LINK](#)

← Woodland caribou



Captured on 19/08/2018



Captured on 21/

Est. Population 22839



Living Area 116 mi²



Data projection by WWF

Wild Life

White Horse
Ruby Mountain
Mt. Soudan

Largehorn Peak

1980

2019

2045

3D

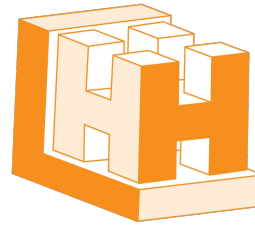


Future Roadmap

	Users	Images	Species trained	Global Analytics	
2020	1000	20M	900	2	Basic analytics Desktop solution Publications
2021	3000	80M	1200	4	Advanced analytics Video support Automated reports Zooniverse integration
2022	6000	200M	1500	6	Video, acoustic support Interactive learning dashboard Mobile app
By 2025	Catalyze evidence-based wildlife management in 2,000 of the world's largest protected areas through real-time data on wildlife distribution and abundance.				

Supporters

GORDON AND BETTY
MOORE
FOUNDATION



LYDA HILL

Google

Thank you!

For more information, visit or contact us at:

<https://wildlifeinsights.org>

info@wildlifeinsights.org

Twitter: @wildinsights

Facebook: @wildlifeinsights

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