

# The fate of tropical forests: High resolution global maps of deforestation risk and future forest cover



Ghislain Vieilledent<sup>1,2</sup>, Christelle Vancutsem<sup>1</sup>, and Frédéric Achard<sup>1</sup>

[1] JRC D.1 Bio-economy unit, [2] Cirad UMR AMAP



## 1 Introduction

- Context
- Objectives

## 2 Data

- Historical deforestation
- Explicative variables

## 3 Modelling

- Statistical model
- Software

## 4 Results

- Parameters
- Spatial probability

## 5 Conclusion

- Projections
- Perspectives

## 1 Introduction

- Context
- Objectives

## 2 Data

- Historical deforestation
- Explicative variables

## 3 Modelling

- Statistical model
- Software

## 4 Results

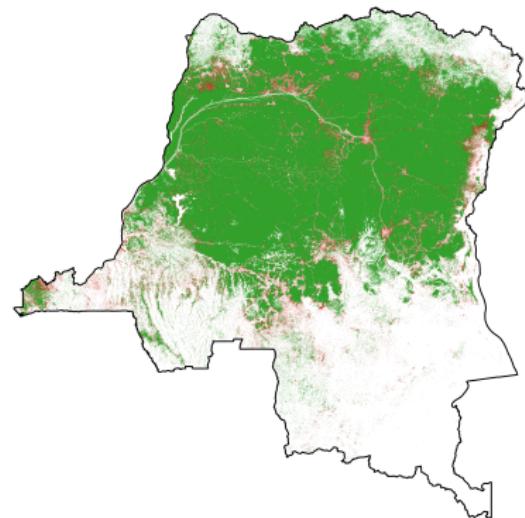
- Parameters
- Spatial probability

## 5 Conclusion

- Projections
- Perspectives

# Long term projections

- Tropical forests shelter most of the terrestrial biodiversity and carbon stocks
- They are currently being deforested at rates close to **1%/yr**

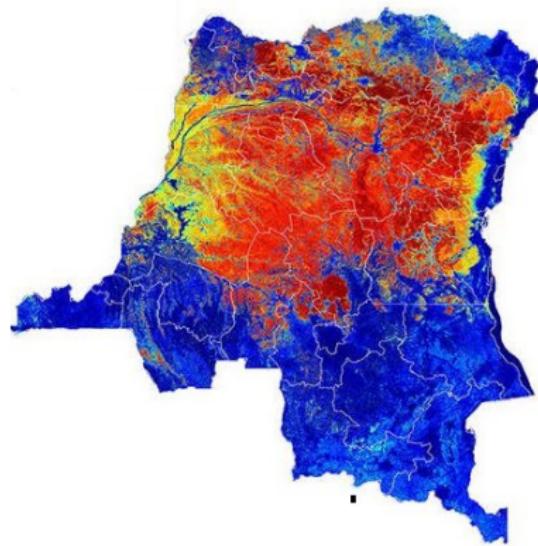


**2005-2015 deforestation in  
Democratic Republic of the Congo**

What happens when you project annual deforestation on the medium or long term (2050-2100) ?

# Spatial projections

- Not all forests are equally threatened
- And biodiversity and forest carbon stocks vary spatially

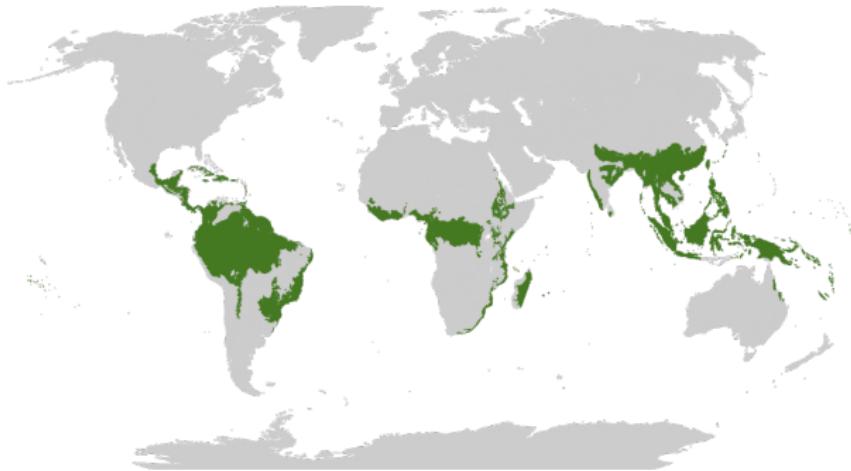


Aboveground biomass in Democratic Republic of the Congo

What are the consequences of long term deforestation for biodiversity and CO<sub>2</sub> emissions ?

# Objectives

- Modelling the deforestation process spatially
- Deriving high-resolution maps of the spatial probability of deforestation
- Projecting forest cover change until 2050 under a business-as-usual scenario
- At the pantropical scale



## 1 Introduction

- Context
- Objectives

## 2 Data

- Historical deforestation
- Explicative variables

## 3 Modelling

- Statistical model
- Software

## 4 Results

- Parameters
- Spatial probability

## 5 Conclusion

- Projections
- Perspectives

# Historical deforestation

- Wall-to-wall map of **tropical moist forest cover change** at 30 m resolution from 1990 to 2018
- Using the 37-years full Landsat satellite archive and Google Earth Engine
- Time-series analysis at the pixel scale using a complex decision tree based on expert knowledge



Introduction  
○○○○

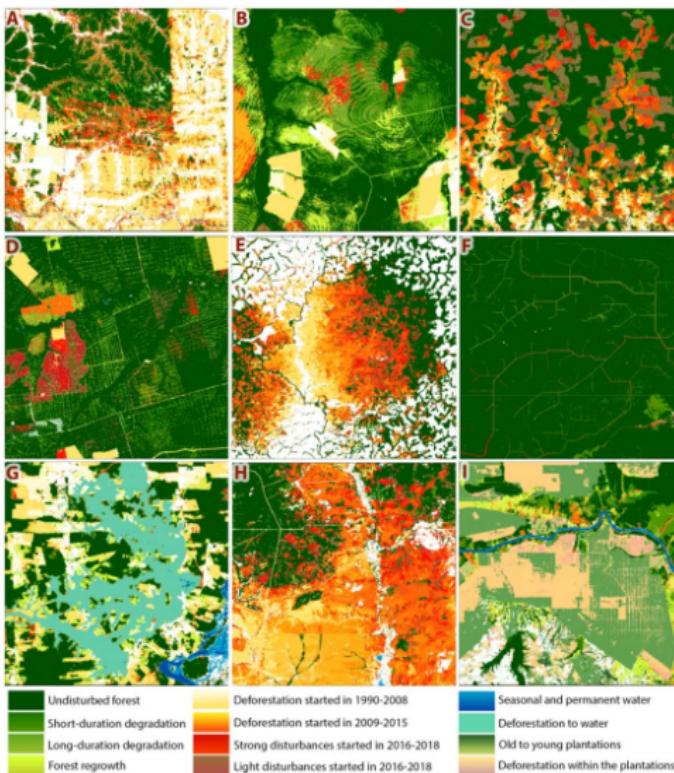
Data  
○○●○○

Modelling  
○○○○

Results  
○○○○○○○○

Conclusion  
○○○○○

# Historical deforestation



# Historical deforestation

- **Vancutsem Ch., F. Achard , J.-F. Pekel , G. Vieilledent, S. Carboni , D. Simonetti , J. Gallego.** Long-term monitoring of the tropical moist forests dynamics reveals unprecedented deforestation rates. Submitted to *Nature Communications*.
- Hansen et al. 2013 : underestimated deforestation rates in Africa (small scale mosaic deforestation)
- Response variable : deforestation on 2005-2015

# Explicative variables

- Variable types : **landscape, accessibility, protection status**

Product	Source	Variable derived	Unit	Resolution (m)
Deforestation maps (2005-2015)	Vancutsem et al. (1)	forest/non-forest	—	30
		distance to forest edge	m	30
		distance to previous deforestation	m	30
Digital Elevation Model	SRTM v4.1 CSI-CIAR (2)	altitude	m	90
Highways	OSM - Geofabrik (3)	slope	°	90
Places		distance to roads	m	150
Waterways	WDPA (4)	distance to towns	m	150
Protected areas		distance to river	m	150
		presence of protected area	—	30

- (1) Vancutsem et al., (2) <http://srtm.csi.cgiar.org>,  
(3) <http://www.geofabrik.de>, (4) <http://protectedplanet.net>

1 Introduction

- Context
- Objectives

2 Data

- Historical deforestation
- Explicative variables

3 Modelling

- Statistical model
- Software

4 Results

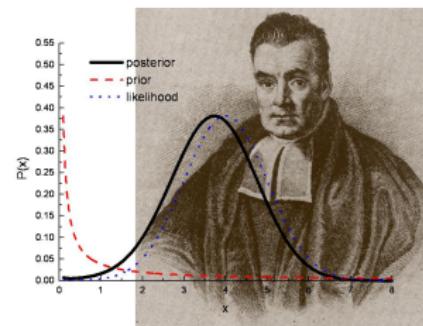
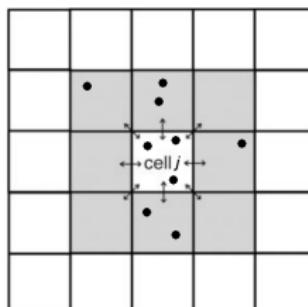
- Parameters
- Spatial probability

5 Conclusion

- Projections
- Perspectives

# Statistical model

- $Y_{ij} \in \{0, 1\} \sim \text{Bernoulli}(\theta_{ij})$
- $\text{logit}(\theta_{ij}) = X_i \beta + \rho_j$
- Autocorrelated spatial random effects  $\rho_j$  (10 km) to account for **unmeasured** or **unmeasurable** factors : population density, soil type, geographical barriers, law enforcement locally
- Structure spatially the **residual variability** that is not explained by the model's variables
- Hierarchical Bayesian framework



Introduction  
○○○○

Data  
○○○○○

Modelling  
○○●○

Results  
○○○○○○○○

Conclusion  
○○○○

# Statistical model

- One model per country
- 40,000 sample points (balanced sampling deforested/non-deforested areas)
- Variable selection (statistical significance + process coherence)

# Software

The screenshot shows the GitHub repository page for 'forestatrisk'. The repository has 257 commits, 1 branch, 1 release, 1 environment, 1 contributor, and is licensed under GPL-3.0. The last commit was made 7 days ago. The repository has 8 issues, 0 pull requests, 0 projects, and a wiki.

Branch: master ▾ New pull request Create new file Upload files Find File Clone or download ▾

File	Action	Time
ghislainv/New tuto	Latest commit cd54275	7 days ago
C	Update	last month
docs	New tuto	7 days ago
forestatrisk	urllib for Python3	last month

- forestatrisk Python package :  
<https://github.com/ghislainv/forestatrisk>
- Rasters processed by chunks : high resolution (30 m, large spatial scale)
- Fast, without memory issues
- Parallel computation : one node per country

## 1 Introduction

- Context
- Objectives

## 2 Data

- Historical deforestation
- Explicative variables

## 3 Modelling

- Statistical model
- Software

## 4 Results

- Parameters
- Spatial probability

## 5 Conclusion

- Projections
- Perspectives

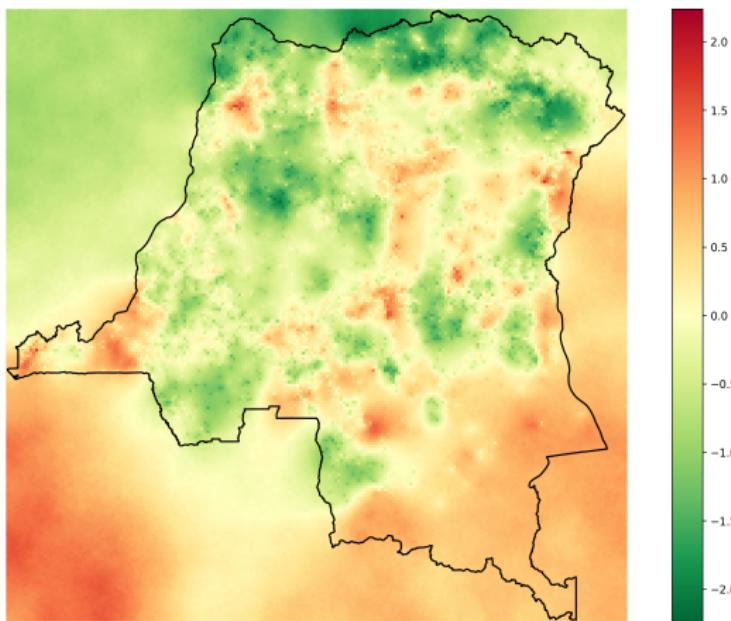
# Parameters

Parameter values :  $\beta$  and variance  $V_p$  of the spatial random effects.

Binomial logistic regression with iCAR process				
Model: I(1 - fcc23) + trial ~ 1 + C(pa) + scale(slope) + scale(dist_defor) + scale(dist_edge) + scale(dist_road) + scale(dist_town) + scale(dist_river) + cell				
Posterior:				
	Mean	Std	CI_low	CI_high
Intercept	-4.64	0.155	-4.92	-4.37
C(pa)[T.1.0]	-0.206	0.101	-0.402	-0.00777
scale(slope)	-0.0505	0.028	-0.113	0.00411
scale(dist_defor)	-5.64	0.304	-6.21	-5.08
scale(dist_edge)	-7.19	0.315	-7.76	-6.54
scale(dist_road)	-0.22	0.0416	-0.303	-0.14
scale(dist_town)	-0.171	0.042	-0.258	-0.0922
scale(dist_river)	-0.0664	0.0311	-0.124	0.00367
Vrho	2.7	0.304	2.32	3.39
Deviance	1.25e+04	89.6	1.23e+04	1.27e+04

- Set of parameters for each country.
- Each effect can be easily interpreted.
- Effects can be compared between countries (efficiency of the protected areas, effect of road infrastructures).

# Spatial random effects



Then interpolated at 1km.

Introduction  
○○○○

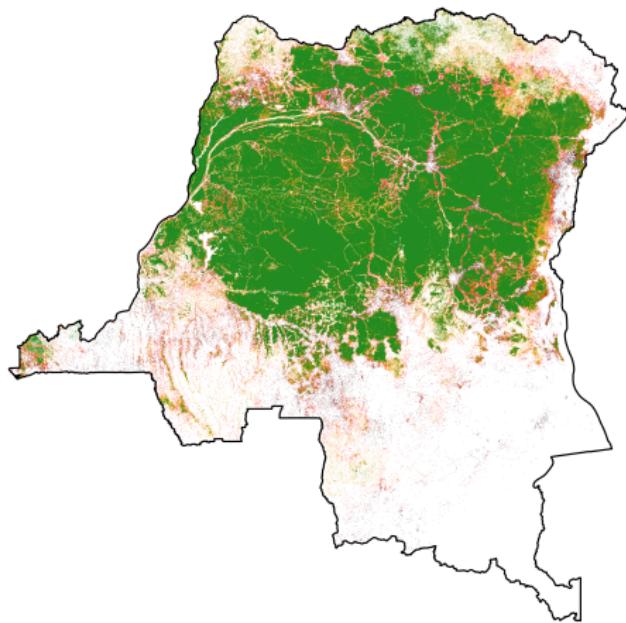
Data  
○○○○○

Modelling  
○○○○

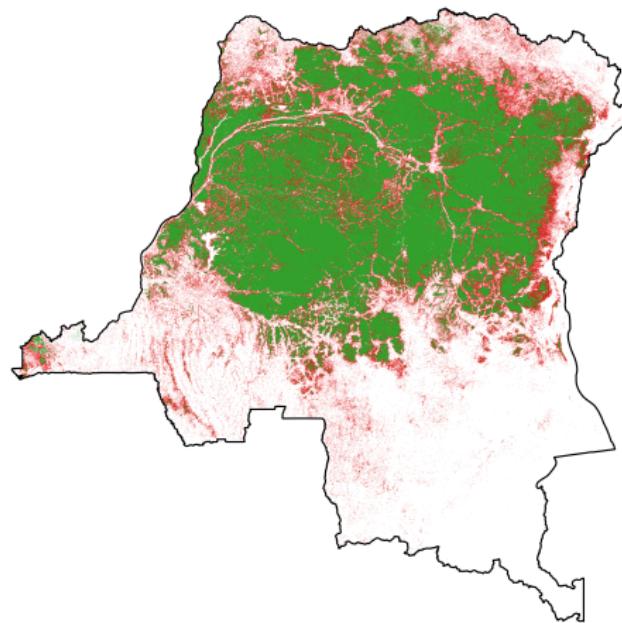
Results  
○○○●○○○○

Conclusion  
○○○○○

# Spatial probability of deforestation



## Future forest cover



Projected forest cover change in **2015-2050** under a business-as-usual scenario.

BAU : historical deforestation (ha/yr) observed on **2005-2015**.

Introduction  
○○○○

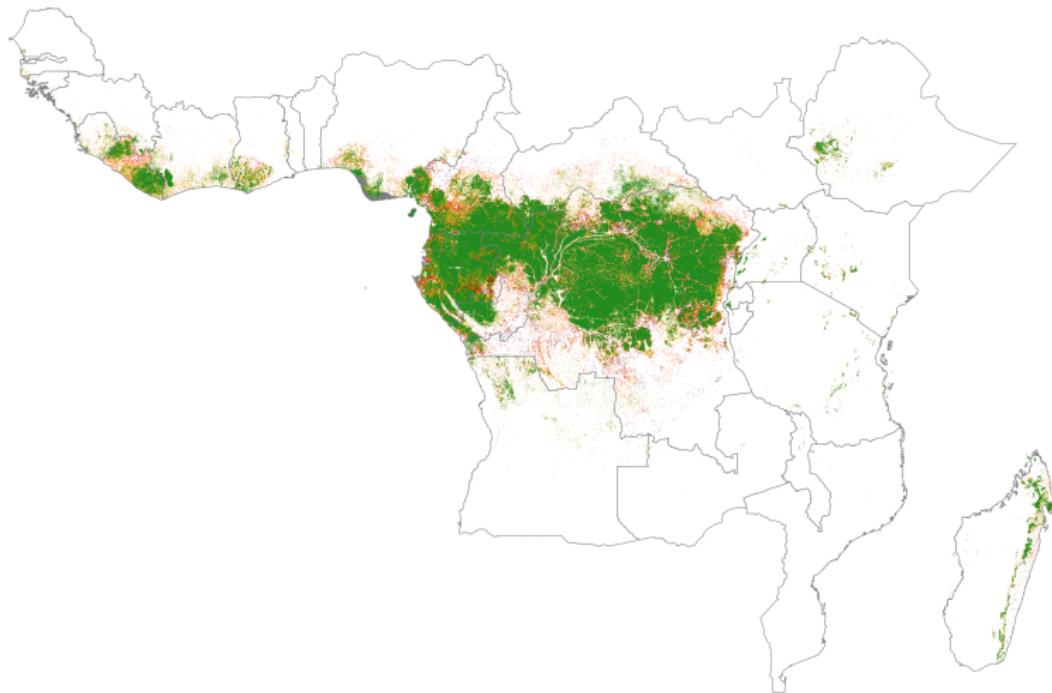
Data  
○○○○○

Modelling  
○○○○

Results  
○○○○●○○○

Conclusion  
○○○○○

# African continent



Spatial probability of deforestation.

Introduction  
○○○○

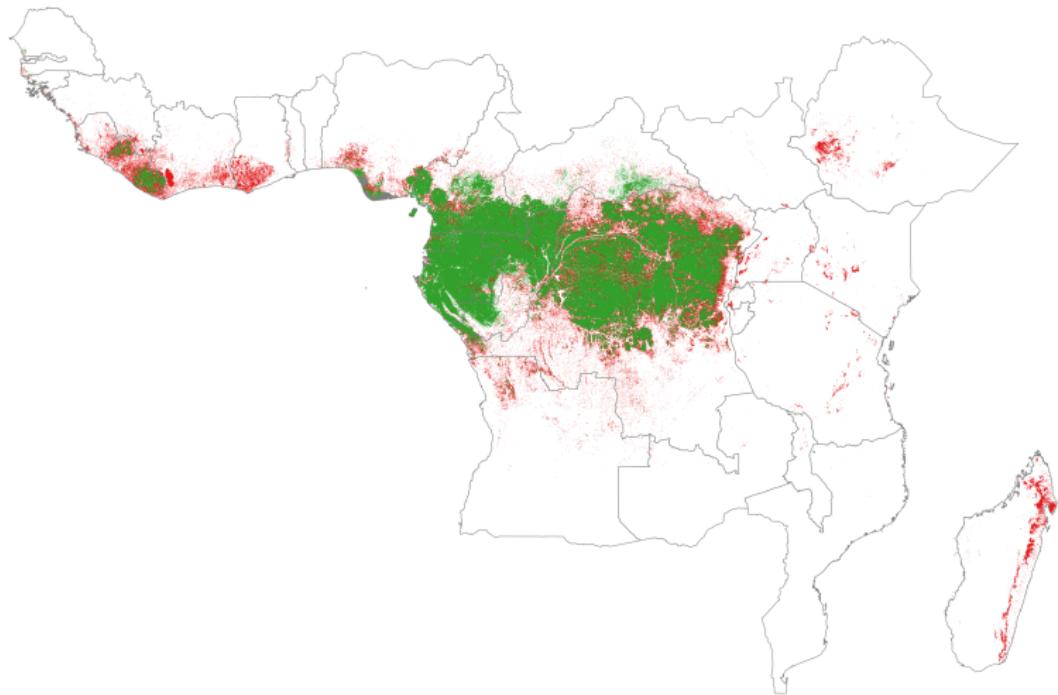
Data  
○○○○○

Modelling  
○○○○

Results  
○○○○○●○○

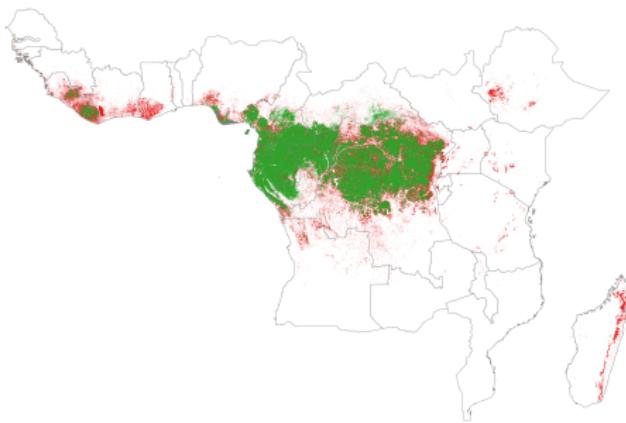
Conclusion  
○○○○○

# African continent



Forest cover change in **2015-2050**, BAU scenario **2005-2015**.

# African continent



- **No more moist forests in 2050** : West-African countries except Liberia, East-African countries including Madagascar
- **Remaining forest block** : Congo, Gabon, Equatorial-Guinea, Cameroon
- **Highly fragmented forest** : Democratic Republic of the Congo
- **Two blocks of forest on both sides of the Congo River**

1 Introduction

- Context
- Objectives

2 Data

- Historical deforestation
- Explicative variables

3 Modelling

- Statistical model
- Software

4 Results

- Parameters
- Spatial probability

5 Conclusion

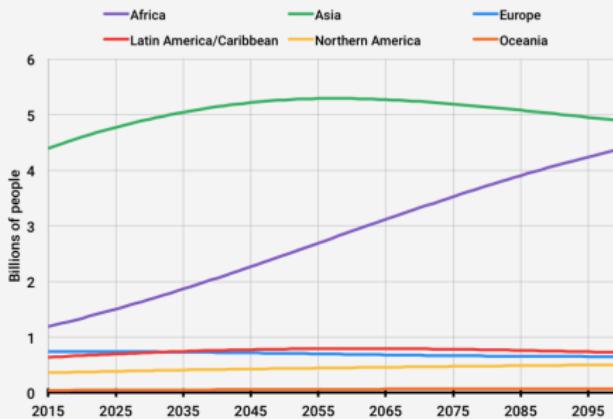
- Projections
- Perspectives

# Projections

- Fake and unrealistic scenario ?

- Simple BAU scenario **driven by observations** (deforestation on 2005-2015)
- Optimistic scenario** : no effect of the demographic growth in Africa in the model

Population projections, 2015-2100



# Projections

- **Projections taken seriously** by the European Commission
- On 23 July 2019, the European Commission presented a **set of actions** to protect and restore the world's forests. [EC Communication](#).

**By 2050, business as usual will wipe out pristine tropical moist forests the size of more than half the EU**

JUL  
26  
2019  
Unless we take action,  
pristine tropical moist forests  
more than half the area of the  
EU will disappear by 2050 –  
new EU data shows

By 2024 the untouched pristine moist forests in their original condition will disappear from Ivory Coast; by 2040, they will be gone from Madagascar, Angola and India, if the current rates of deforestation and forest degradation are kept, according to a new map by the Joint Research Centre (JRC).



By 2050, business as usual will wipe out pristine tropical forests the size of more than half the EU  
©Thekopmylife, AdobeStock

# Perspectives

- Maps can be used to identify **high-priority conservation areas** (biodiversity, carbon, etc.)
- **Additional scenario** taking into account the demographic growth
- Extending the work to the **entire tropics**
- **Spreading the word** to (hopefully) have impact on policies





... Thank you for your attention ...  
<https://ecology.ghislainv.fr/presentations>  
 @ghislainv  
ghislain.vieilledent@cirad.fr

