



## Modelling and forecasting tropical deforestation: advances and perspectives



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## 1 Deforestation and demography in Africa

- Model
- Results
- Valorisation

## 2 `forestatrisk` Python module

- Functionalities
- Improvements
- Model performance

## 3 Forecasting spatial deforestation spatially

- Pantropical scale
- Computational challenge
- Scenario choice

## 4 Perspectives

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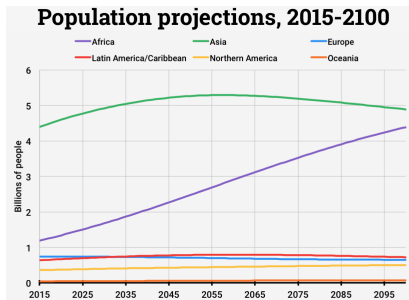
- Computational challenge

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# Model

- The fate of African tropical forests
- Associated to demographic explosion
- $\log D = \beta_0 + \beta_1 \log F + \beta_2 \log P$
- Data on deforestation :
  1. JRC : 1990-2000-2010
  2. GFC :  
2000-2005-2010-2015
- Projection of forest cover in 2050, 2100



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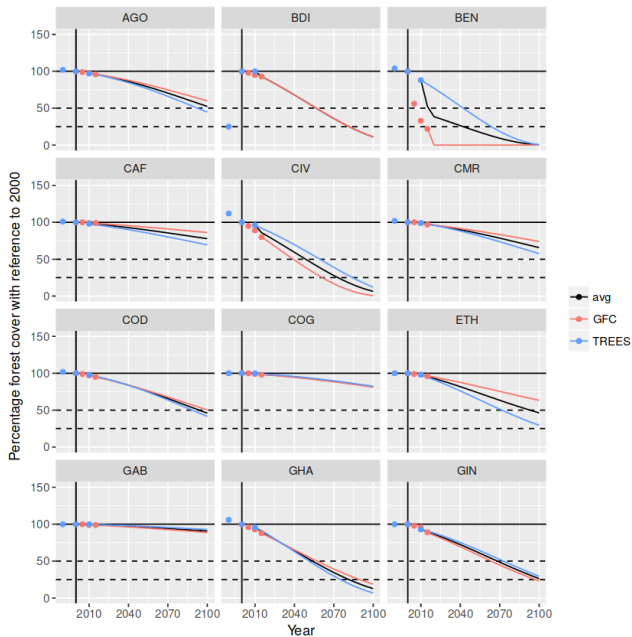
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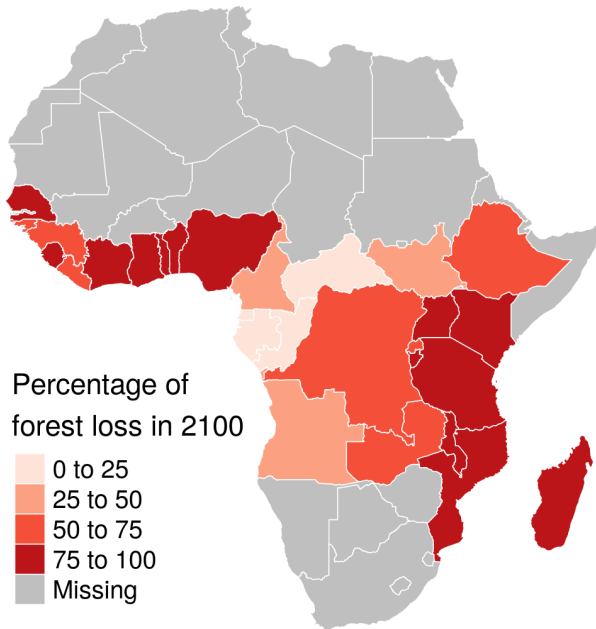
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# Model averaging per country





## Percentage of forest loss 21st century



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# Valorisation

- Scientific article
- Integration of Roadless data on deforestation ?
- Use of the results for future deforestation scenario in Africa
- Predictions in percentage of forest loss :  $\sim$  independent of forest definition

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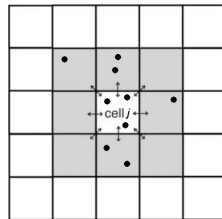
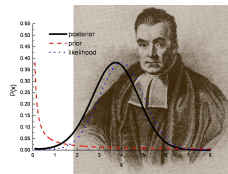
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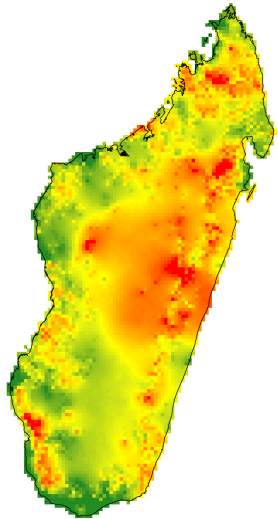
# forestatrisk Python module specifications

- Spatial probability of deforestation
- $\text{logit}(\theta_i) = f(\text{spatial factors}_i) + \rho_j$
- Factors : accessibility (dist. towns, roads, villages), landscape (dist. forest edge), land-tenure (protected areas)
- $\rho_j$  : spatial random effect
- $\{\text{\href{https://github.com/ghislainv/forestatrisk}}\}$



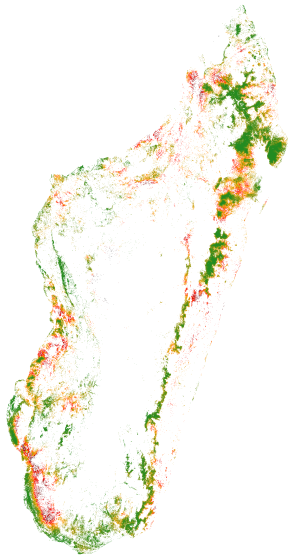
## Spatial random effects

- Hotspots of deforestation
- Not explained by the fixed env. factors



# Spatial probability of deforestation

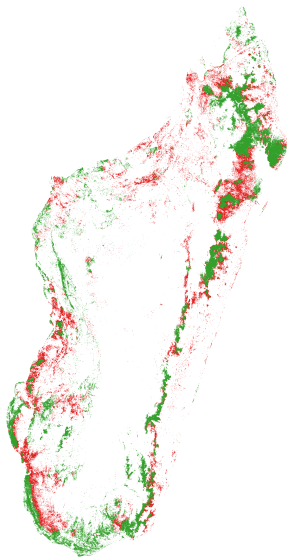
- Computed at 30 m resolution
- Greener : lower probability
- Darker red : higher probability





# Future forest cover

- green : residual forest in 2050
- red : deforested area  
2010-2050



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# Improvements

- Python 2.7 and Python 3.x compatible
- Spatial random effects limited to country border
- Set of new functions for model validation
- Can be used from R with `reticulate`



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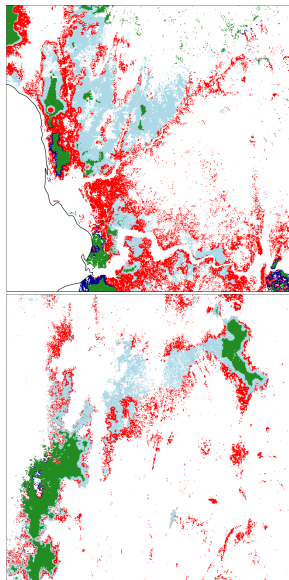
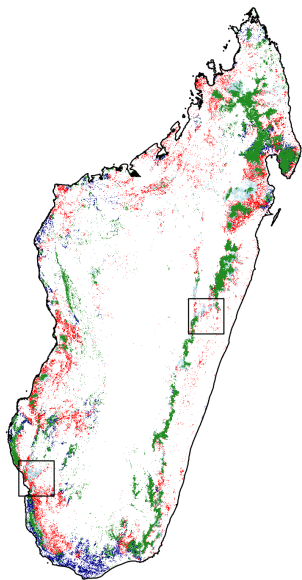
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## Comparing predictions with GLM



## Model performance

Performance estimated on an independent data-set of 20,000 points

model	Deviance	OA	Kappa
null	0	50	0
GLM	8	62	24
iCAR	30	79	58
full	100	100	100

## Model validation

- Map of probability of deforestation in 2010 + known deforested area on 2010-2014
- Observed vs. projected deforestation on 2010-2014
- Area deforested in  $10 \times 10$  km areas
- Pearson's correlation

model	Cor
GLM	12
iCAR	31

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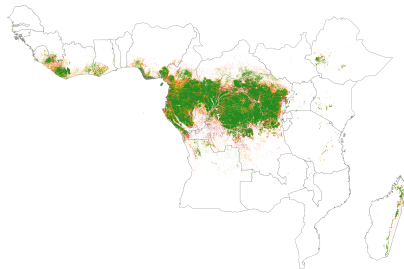
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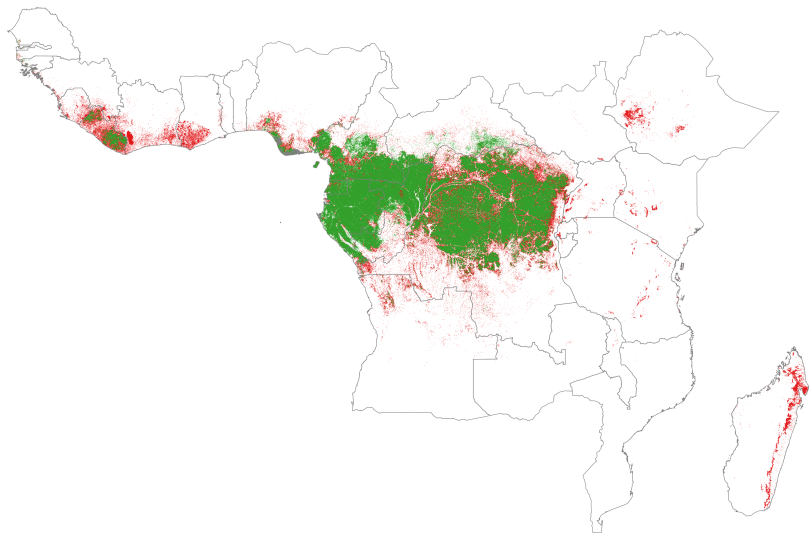
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# Africa

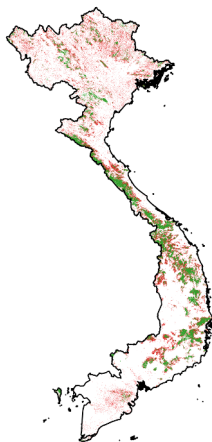
- Map of deforestation probability in 2015
- Future forest cover in 2050, 2100





# Asia

- 11 countries in tropical Asia
- Including MMR, THA, KHM, LAO, VNM (ReCaREDD focus countries)
- Ex. Vietnam in 2050 (half current deforestation rate)



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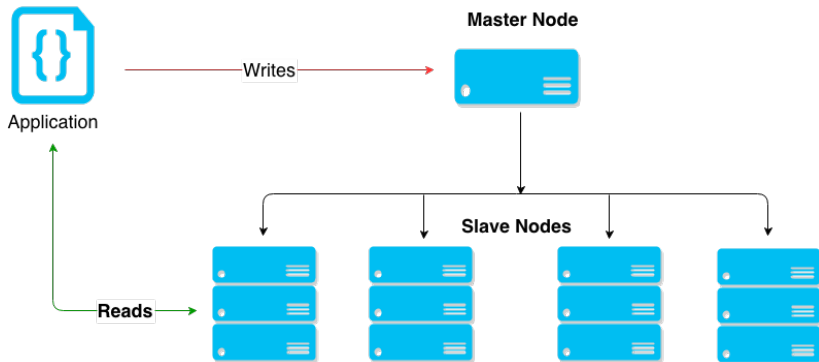
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# Computational challenge

- Use of Google Cloud Computing ( $\neq$  GEE)
- Cluster of small machines with some cores
- Parallelization : one country per machine



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## Scenario choice

- “business-as-usual”
- demographic growth for African country
- mitigation scenario (50% deforestation)



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# Perspectives

## 1. Deforestation-demography study

- consolidate the results
- publish a first paper

## 2. forestatrisk Python module

- polish the code
- publish a methodological paper advocating our model choice

## 3. Pantropical map

- re-run computation with new version of the model
- extend projections to the pantropical world
- produce maps for alternative scenario of intensity

... Thank you for attention ...