

Integrated Economic-Environmental Modeling (IEEM) for Evidence-Based Public Policy and Investment Design.

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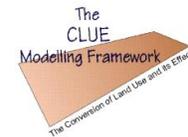
RMGEO Consultants Inc.



**LAND USE LAND COVER
CHANGE MODELING
FOR
IEEM**

LULC MODELING FOR IEEM

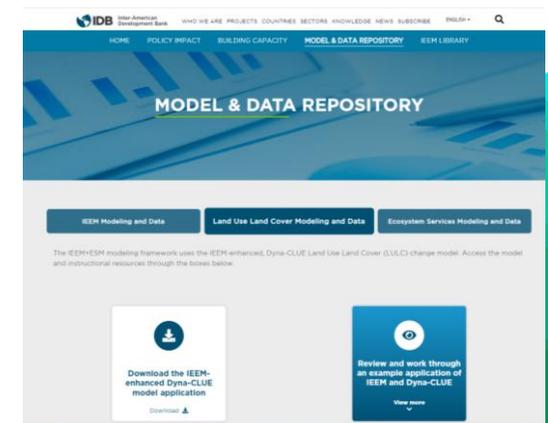
- Basic requirements for IEEM integration:
 1. Can use demand for land as an input (this is derived from IEEM).
 2. Outputs in the form of LULC raster maps.
 3. Can handle multiple regions. **Why?**
 4. Capacity to operate at 'adequate resolution' for ES modeling.
 5. Straightforward to format data and run, good error detection.
- We use the Dynamic Conversion of Land Use and its Effects (Dyna-CLUE) model.
- Applied in 180+ studies in 28 countries globally at varying scales.
- Visit OPEN IEEM for Dyna-CLUE manual and **video tutorial series**.



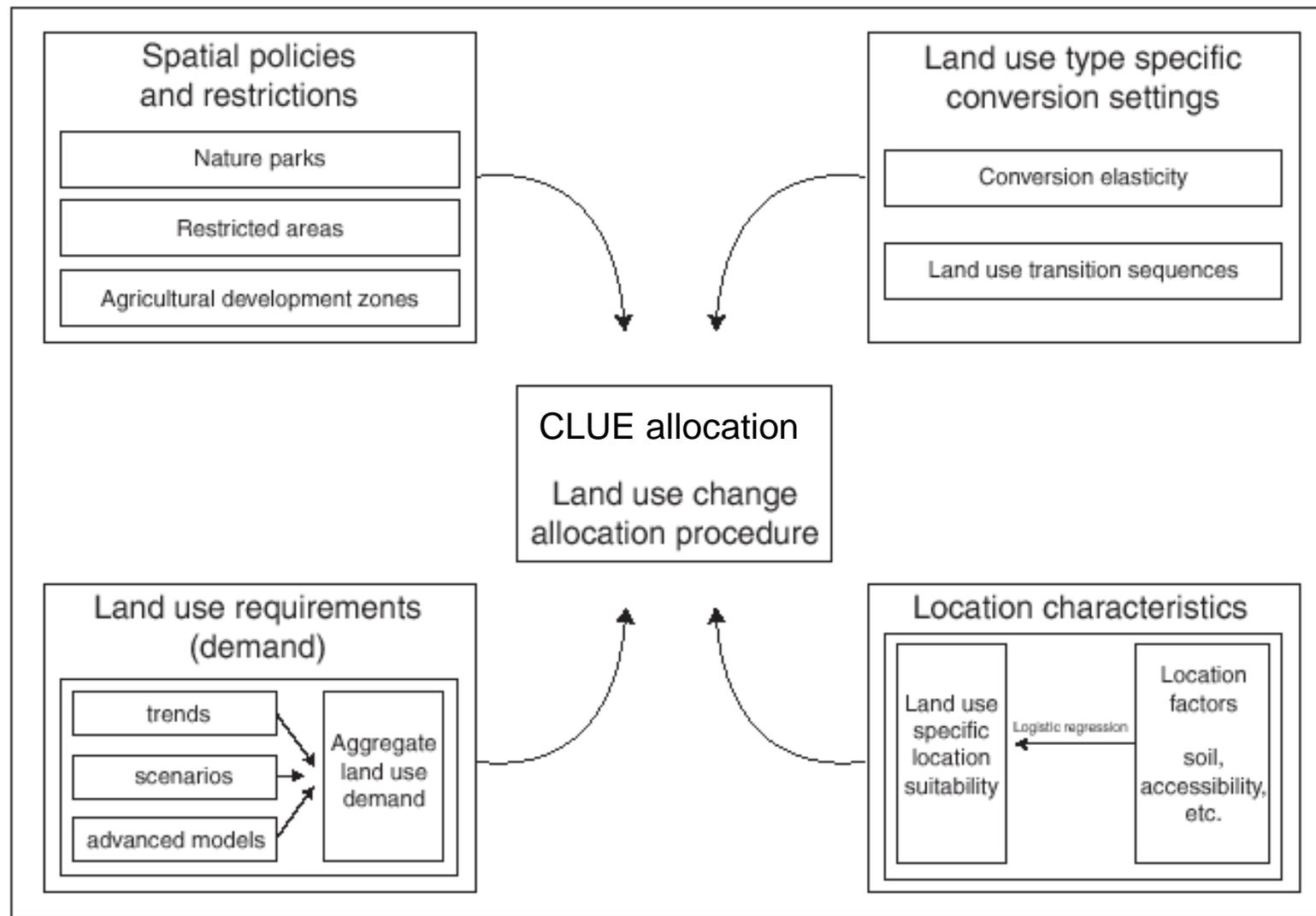
The CLUE model
[Article](#)
[More info and model.](#)



The CLUMondo land systems model
[Article](#)
[More info and model.](#)

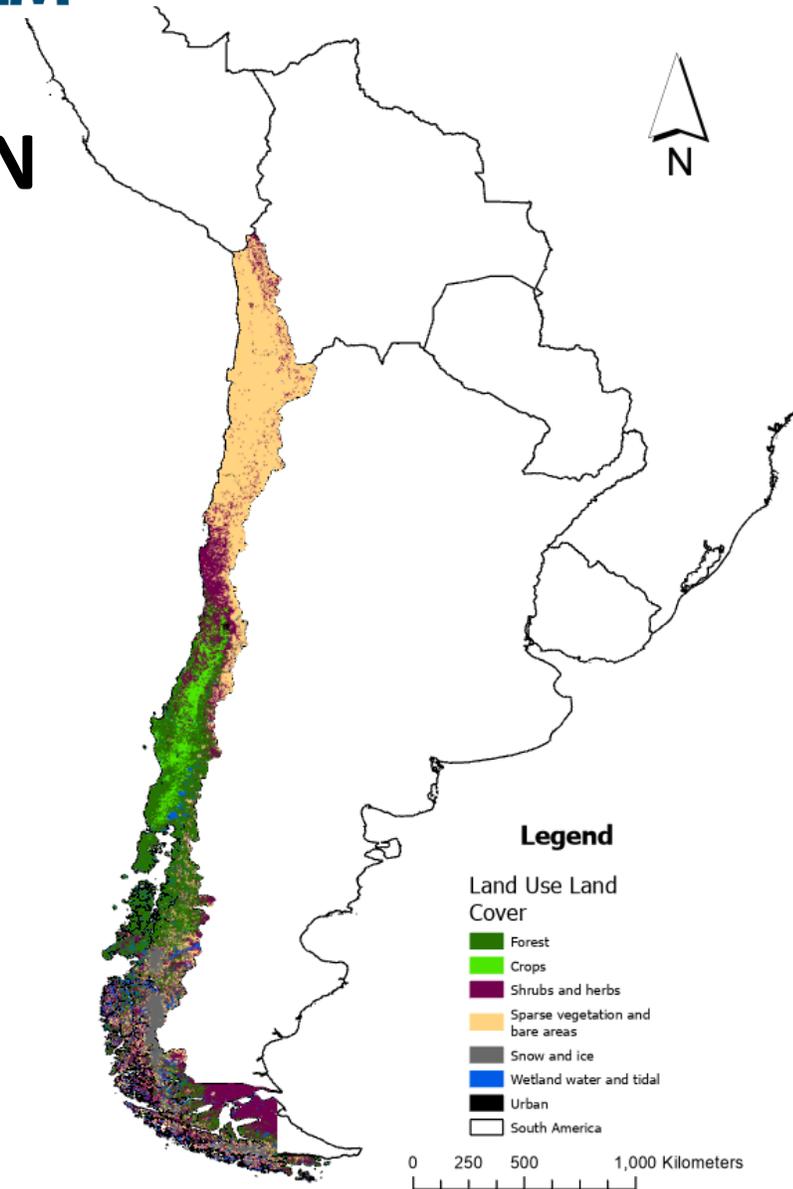


DYNA-CLUE OVERVIEW



DYNA-CLUE DATA PREPARATION

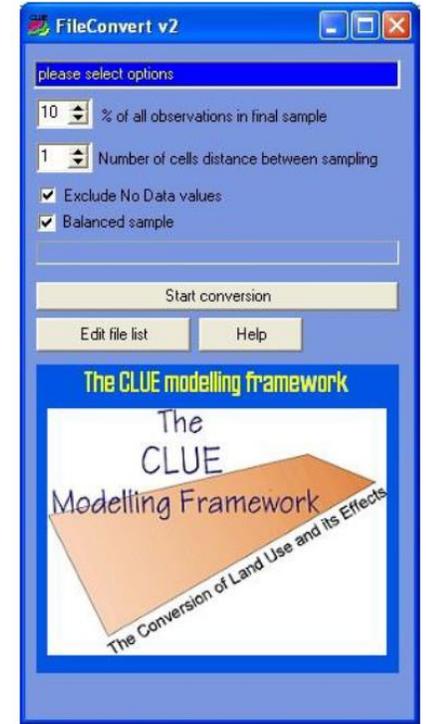
- LULC map (national or globally sourced). 7 to 10 classes, some static.
- Regions: Amazon tipping point: 26; Habitat Banking/Colombia: 30.
- Base Map: Copernicus 2019 Global Land Service, 100m; 23 classes reclassified to 7.
- Level of disaggregation depends on policy question.
- IEEM baseline will use LULC map initial areas (cross check census data); includes forest, crops (different disaggregations) and grassland.



Projection: SIRGAS-Chile 2021 UTM Zone 19S.
Based on 20 Copernicus Global Land Service Land Cover.

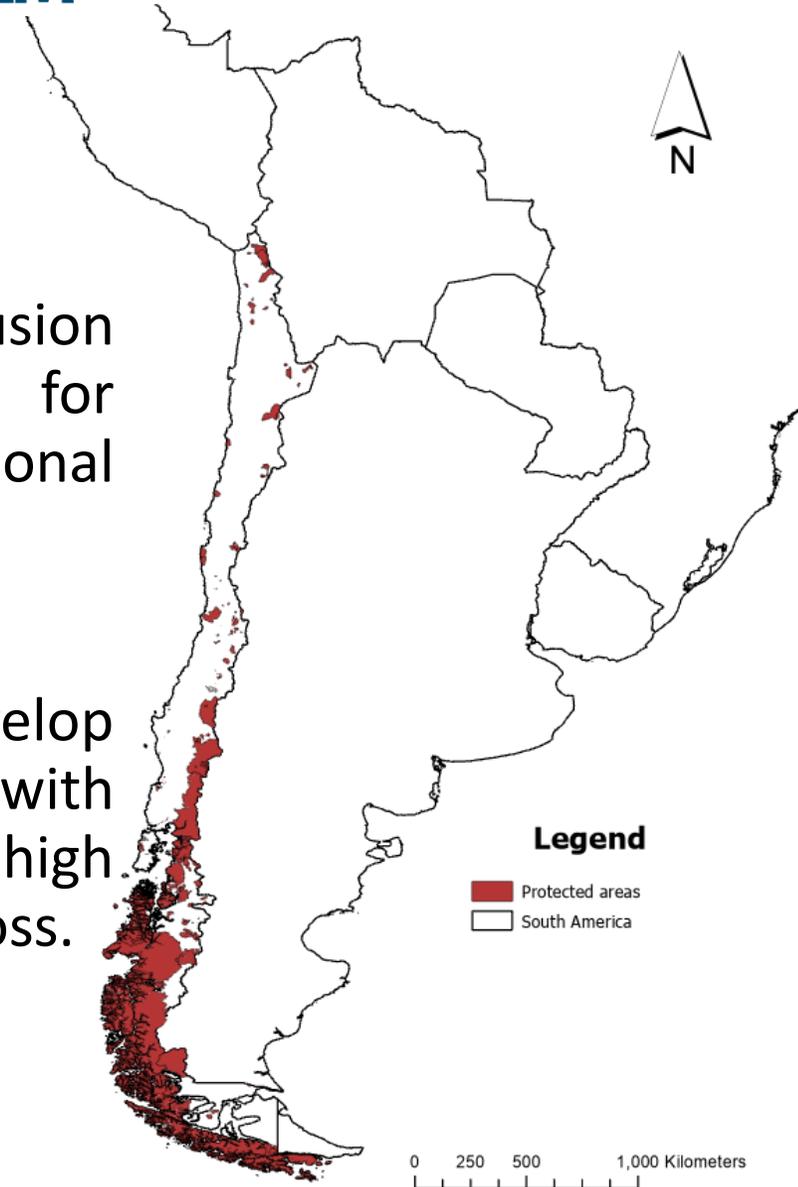
DYNA-CLUE FILE PREPARATION

- Create mask of area of interest: region file with exclusion areas. Apply regional mask to all ascii spatial inputs to ensure same dimensions.
- Create binary presence/absence LULC class maps.
- Extract driving factors.
- Use conversion tool to translate spatial data into a tabular format for import in statistical packages. One table for each binary LULC map and all drivers.
- Run stepwise regression in statistical package (Stata, R, SPSS, by hand).



EXCLUSION AREAS

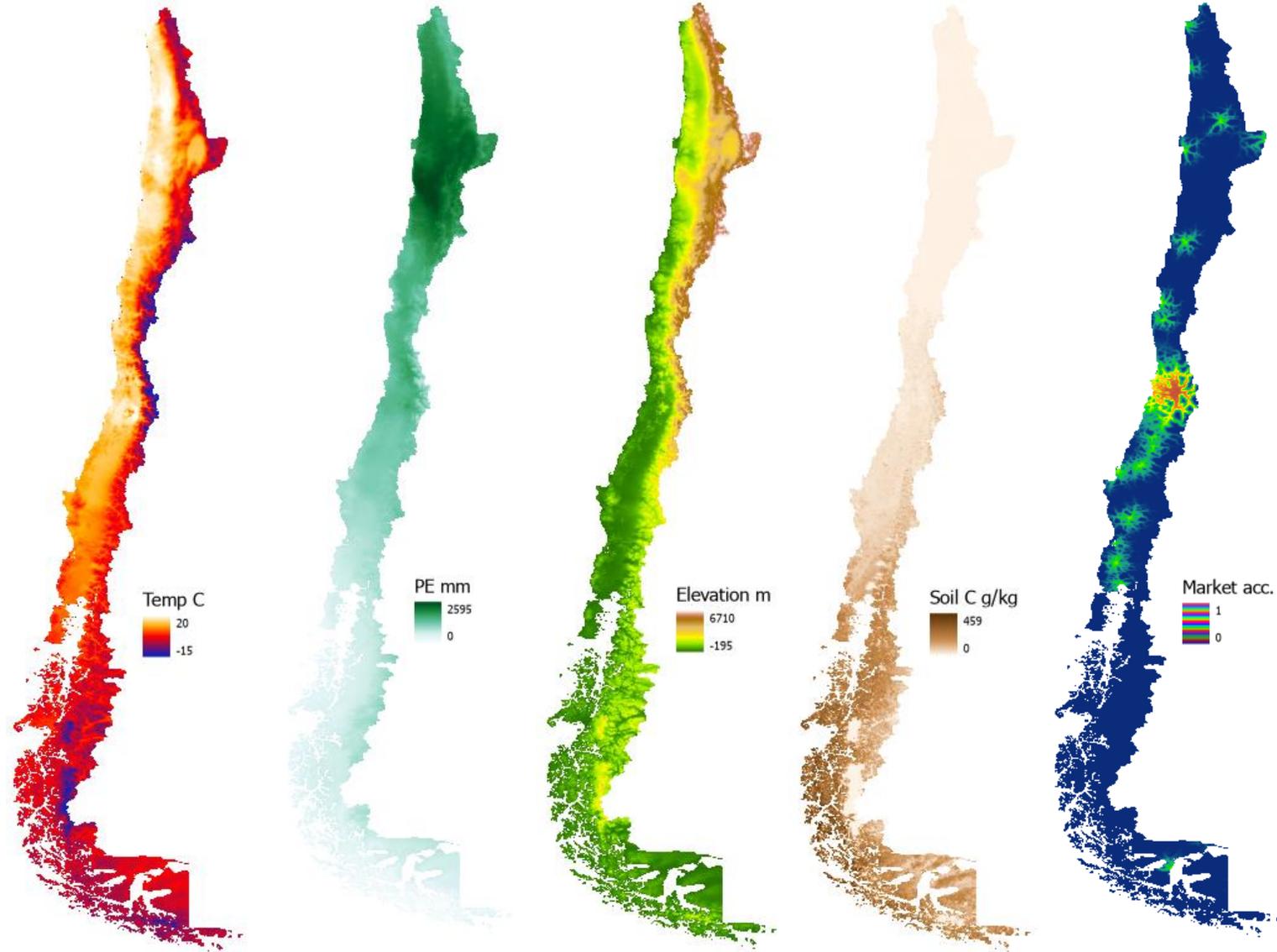
- Protected areas and other areas. Exclusion renders these areas ineligible for conversion. Protected areas from national source or 'Protected Planet' database.
- Using ES model outputs to develop exclusion layers, for example, areas with high pollinator abundance or high susceptibility for nutrient or sediment loss.



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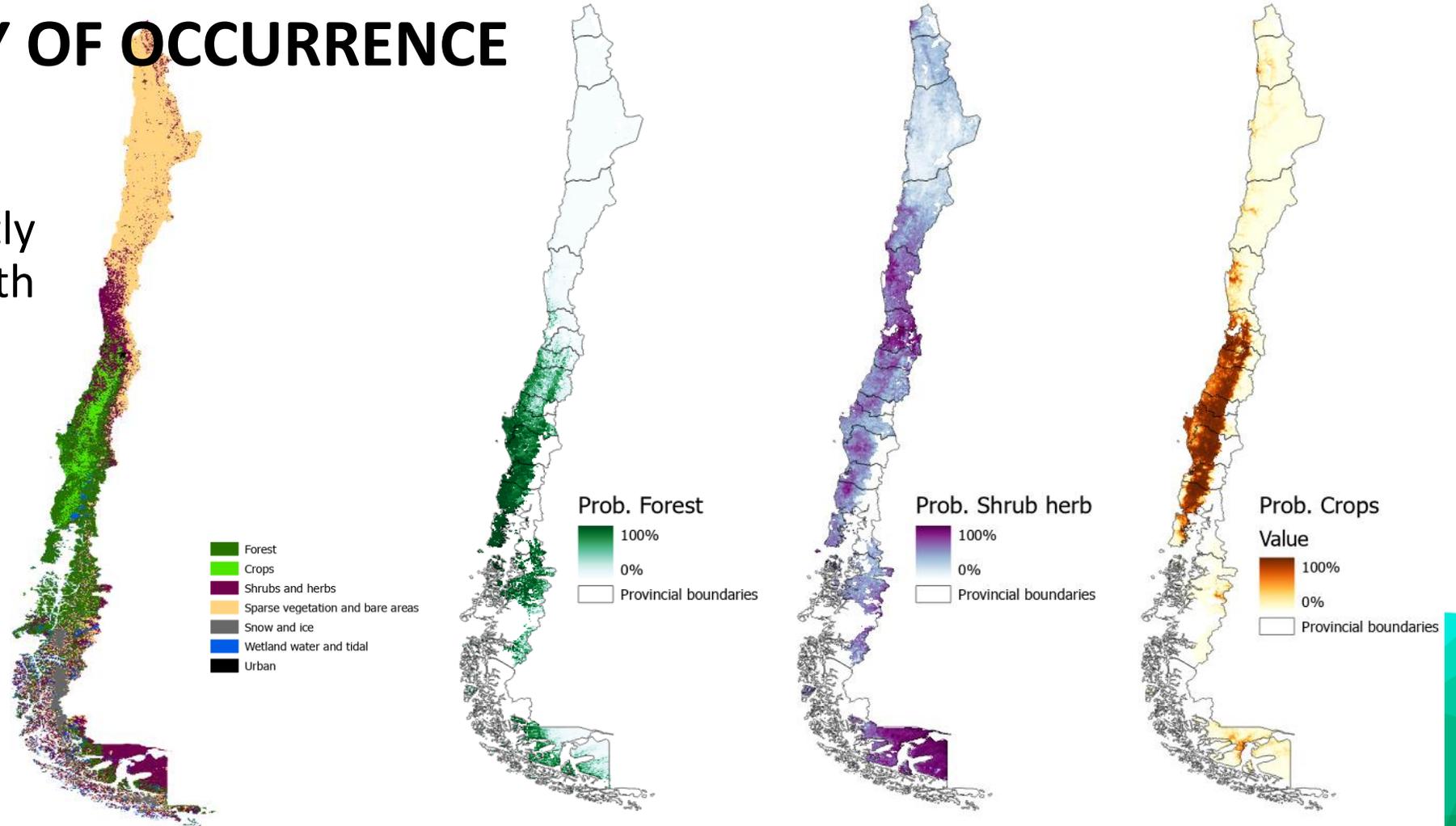
LULC DRIVING FACTORS

- 15 driving factors (climate, topography, soils, socioeconomic factors).
- Apply mask to (global) databases.
- LULC Datapacket; contains all spatial layers and example application; reuse.



PROBABILITY OF OCCURRENCE

- Probability usually mostly aligned with current LULC.



DYNA-CLUE FILE PREPARATION

- Allow matrix.
- Allocation file (regression results).
- Main parameters file includes:

-dimensions of project (~3,000 rows or columns max), number of LULC classes, driving factors; conversion resistance; error terms

| | FUTURE | Forest | Crops | Shrubs herb | Sparse veg | Snow and ice | Wetland, water | Urban |
|----------------|--------|--------|-------|-------------|------------|--------------|----------------|-------|
| PRESENT | | 0 | 1 | 2 | 3 | 4 | 5 | 6 |
| Forest | 0 | 1 | 1 | 0 | 0 | 0 | 0 | 0 |
| Crops | 1 | 0 | 1 | 0 | 0 | 0 | 0 | 0 |
| Shrubs herb | 2 | 1 | 0 | 1 | 0 | 0 | 0 | 0 |
| Sparse veg | 3 | 0 | 0 | 0 | 1 | 0 | 0 | 0 |
| Snow and ice | 4 | 0 | 0 | 0 | 0 | 1 | 0 | 0 |
| Wetland, water | 5 | 0 | 0 | 0 | 0 | 0 | 1 | 0 |
| Urban | 6 | 0 | 0 | 0 | 0 | 0 | 0 | 1 |

BASE

 Forest Crops Shrubs Sparse veg Snow/ice Wet/water Urban
 0 1 2 3 4 5 6

| Years of simulation | 0 | 1 | 2 | 3 | 4 | 5 | 6 |
|---------------------|------------|-----------|------------|------------|-----------|-----------|---------|
| 2020 Year 0 | 17,728,281 | 4,213,750 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2021 Year 1 | 17,609,241 | 4,332,791 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2022 Year 2 | 17,490,999 | 4,451,032 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2023 Year 3 | 17,373,552 | 4,568,479 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2024 Year 4 | 17,256,894 | 4,685,138 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2025 Year 5 | 17,141,018 | 4,801,013 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2026 Year 6 | 17,025,921 | 4,916,110 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2027 Year 7 | 16,911,597 | 5,030,435 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2028 Year 8 | 16,798,040 | 5,143,991 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2029 Year 9 | 16,685,246 | 5,256,786 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2030 Year 10 | 16,573,209 | 5,368,822 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2031 Year 11 | 16,461,924 | 5,480,107 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2032 Year 12 | 16,351,387 | 5,590,644 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2033 Year 13 | 16,241,592 | 5,700,439 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2034 Year 14 | 16,132,534 | 5,809,497 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2035 Year 15 | 16,024,209 | 5,917,823 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2036 Year 16 | 15,916,610 | 6,025,421 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2037 Year 17 | 15,809,735 | 6,132,296 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2038 Year 18 | 15,703,577 | 6,238,454 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2039 Year 19 | 15,598,132 | 6,343,900 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2040 Year 20 | 15,493,394 | 6,448,637 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2041 Year 21 | 15,389,361 | 6,552,671 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2042 Year 22 | 15,286,025 | 6,656,006 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2043 Year 23 | 15,183,384 | 6,758,648 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2044 Year 24 | 15,081,431 | 6,860,600 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2045 Year 25 | 14,980,164 | 6,961,867 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2046 Year 26 | 14,879,576 | 7,062,455 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2047 Year 27 | 14,779,664 | 7,162,367 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2048 Year 28 | 14,680,422 | 7,261,609 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2049 Year 29 | 14,581,847 | 7,360,184 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |
| 2050 Year 30 | 14,483,934 | 7,458,097 | 16,201,563 | 31,173,906 | 2,937,031 | 2,923,750 | 251,875 |

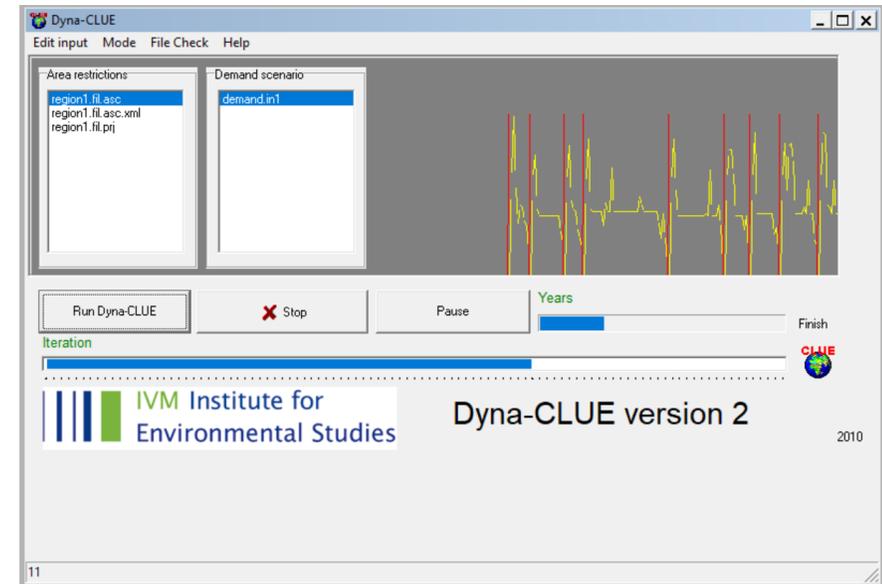
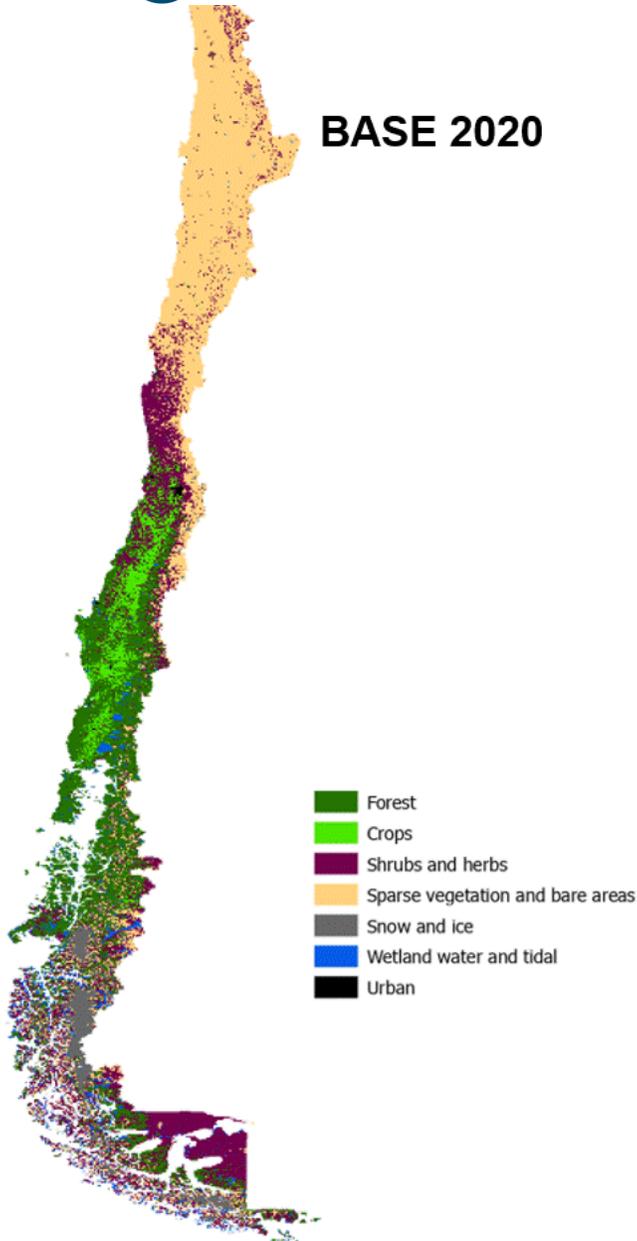
| | | | | | | | |
|--|------------|---------|------------|------------|-----------|---------|--------|
| MODEL RUI | 14490781.2 | 7454375 | 16198437.5 | 31173906.2 | 2937031.2 | 2923750 | 251875 |
| DEVIATION BETWEEN DEMAND AND RESULT | | | | | | | |
| | 0.05% | -0.05% | -0.02% | 0.00% | 0.00% | 0.00% | 0.00% |

DYNA-CLUE FILE PREPARATION

- Demand file for BASE (baseline) and each scenario derived from IEEM.
- Exogenous vs. endogenous land use? Time steps.
- Once application is working, scenarios can be implemented readily (LULC datapackets).

RUN DYNA-CLUE

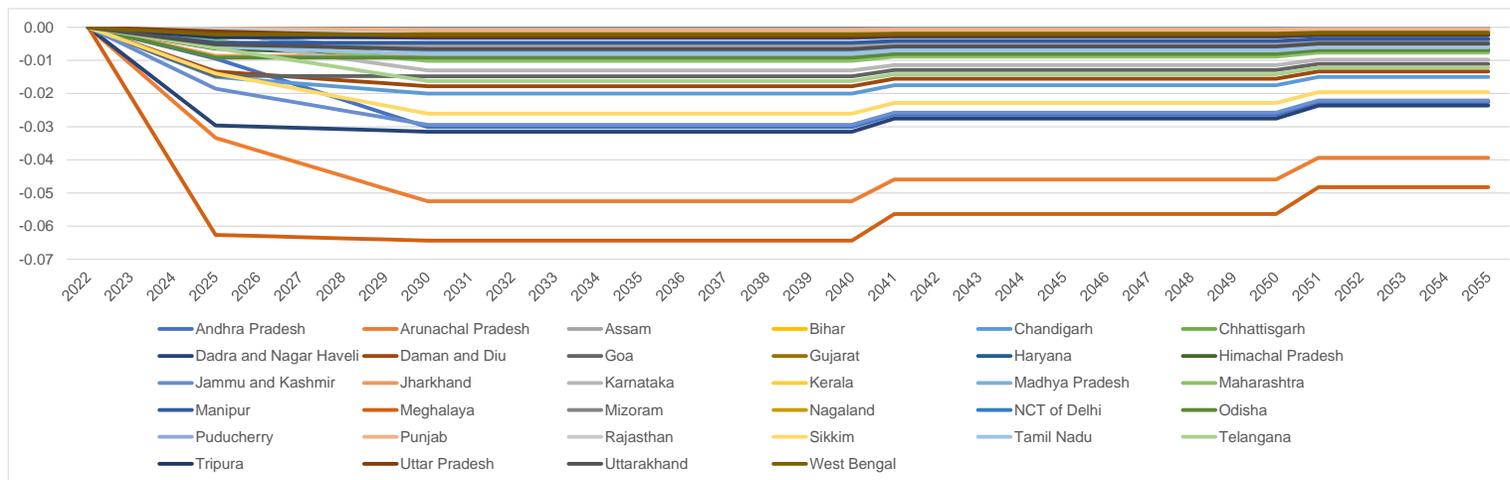
- Baseline projection (deforestation, agricultural land expansion).
- Run scenarios.
- Convert outputs to .tif with projection for use in InVEST.



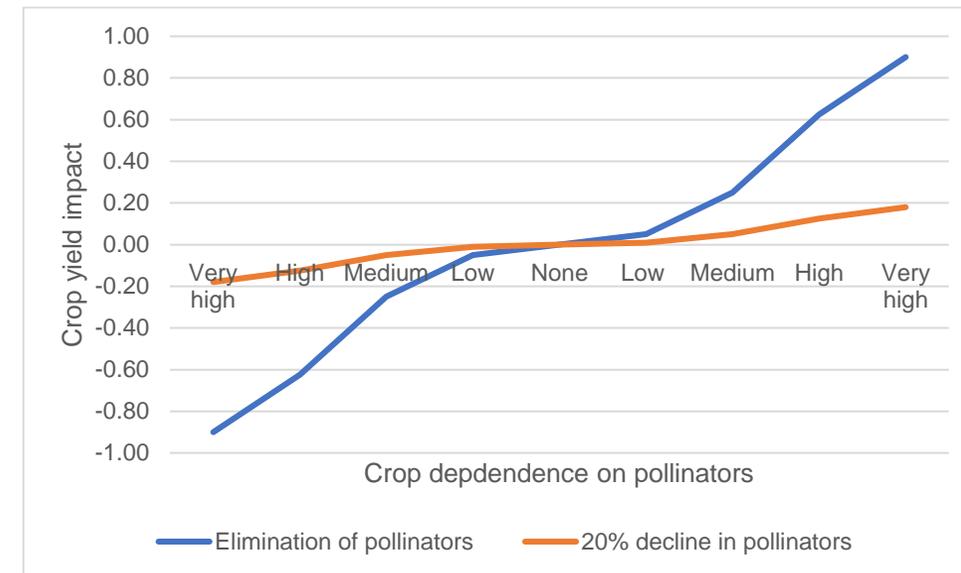
USING LULC MODELING RESULTS

- With Sediment Delivery Ratio and crop pollination models, calculate shocks to implement in IEEM. Single region vs. multiple regions.
- Exogenous vs. endogenous land use; **Brazil CCDR**.

Erosion shock



Crop pollination shock



CONCLUDING REMARKS

- Dyna-CLUE has worked relatively well, long history of applications across the globe.
- Intense data preparation requirements outside of model. Time consuming. Need tools to simplify data preparation/processing. No intuitive error detection.
- Dimensional limit which makes regional modeling at appropriate resolution challenging. Options include aggregation along agroecological zones; independent application by country (country demands for land); lower resolution.
- Funded by the  IDB Inter-American Development Bank IEEM+ESM LULC change model beta version developed. More memory. facilitates integration with IEEM. Enhanced error detection. Coming soon!
- [See an application?](#)

Developing IEEM Modeling Infrastructure and Capacity Around the World.

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