



# Multiobjective stochastic optimisation for long and short-term planning of prepositioning in humanitarian logistics

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*UCM-HUMLOG: Decision aid models for logistics and disaster management (Humanitarian Logistics)*

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1. Disaster Risk Reduction and Humanitarian Logistics
2. A model for prepositioning and warehouse location for preparedness
3. Case study: Mozambique
4. Modelling alternatives
5. Conclusions

# Contents

1. **Disaster Risk Reduction and Humanitarian Logistics**
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# Disaster Risk Reduction

- **DRR:** concept and practice of risk reduction through systematic effort to analyse and reduce the causal factors of disasters
  - Reduction of the exposure to hazards
  - Reduction of the population vulnerability
  - Rational management of the land and the environment
  - *Improvement of preparedness and response to adverse events*
  
- **DISASTER:** Disruption of the normal functioning of a system or community, in a specific geographical area and limited time, which causes a strong impact on people, structures and environment, and **goes beyond local capacity of response.**

## Levels and entities (decision makers)

# 1



### **Local Level (L1)**

Civil organizations and population  
Local agents (police, firefighters...):  
Local civil protection

# 2



### **National/regional Level (L2-L3)**

National civil protection, Army, Other  
governmental agencies, Red Cross and  
Humanitarian NGOs

# 3

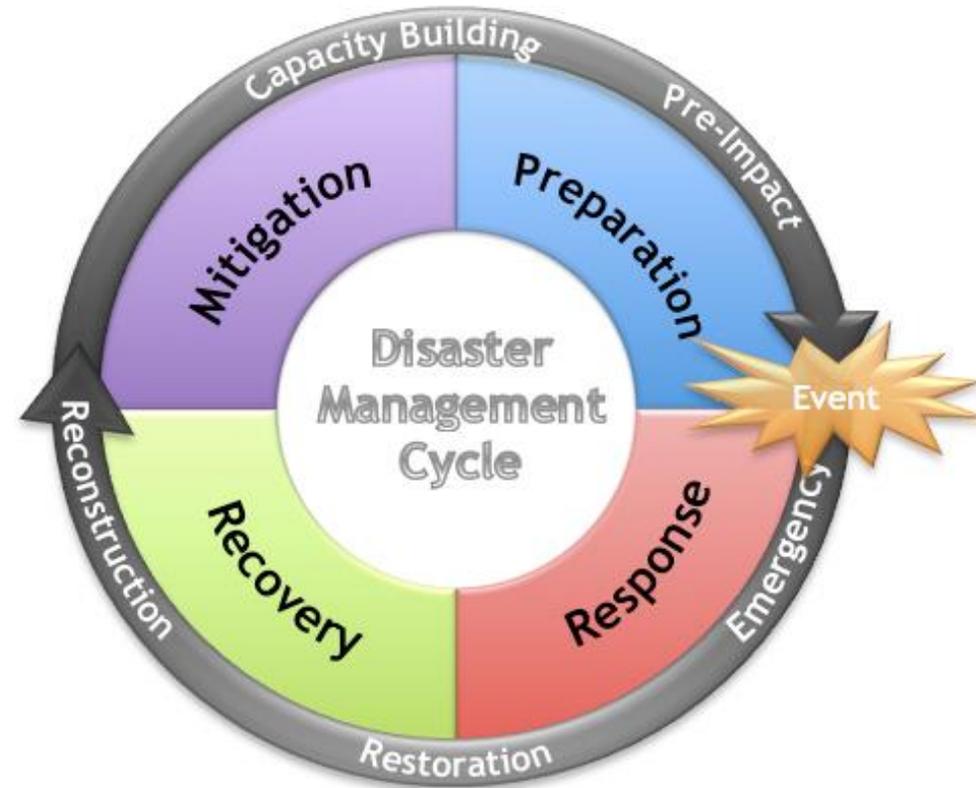


### **International Level (L4-L5)**

Foreign governments (ECHO, USAID)  
IHO.: Red Cross, MSF, ACF, Oxfam, Care,  
World Vision, Save the Children  
UN: OCHA, Agencies (UNICEF, WFP, UNDP,  
UNHCR). IASC (Inter-Agency Standing  
Committee)

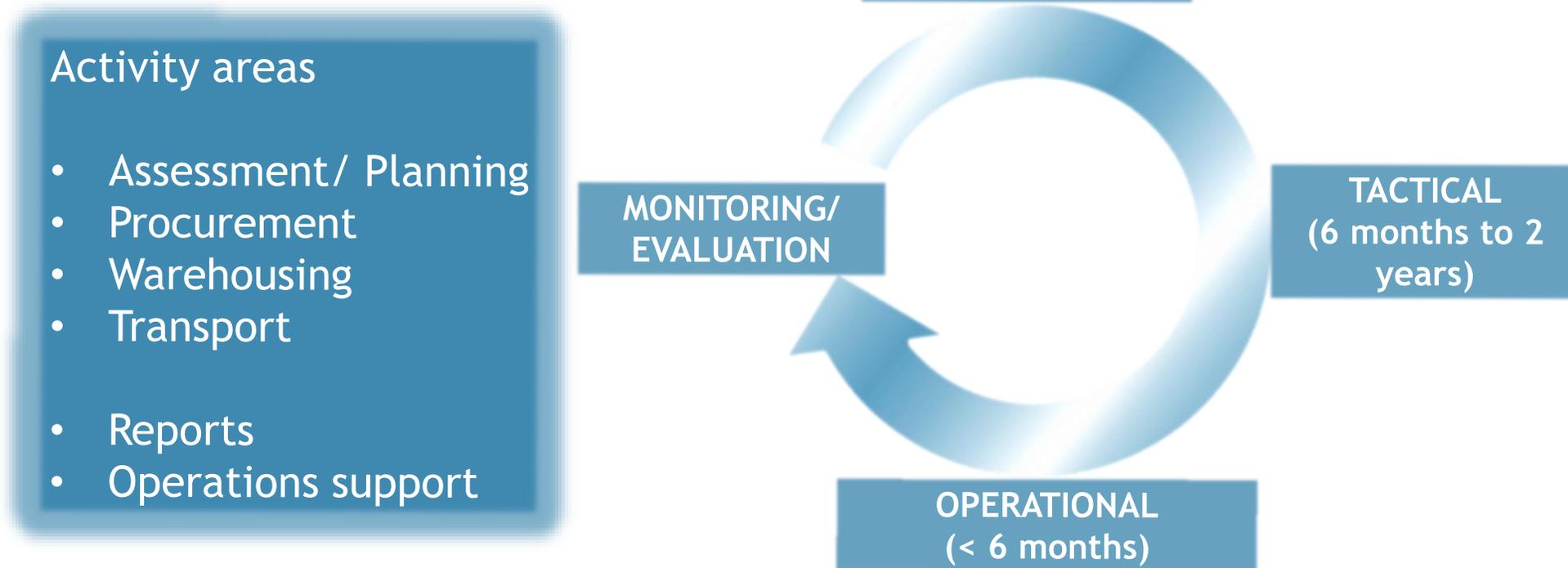
# The disaster cycle

- Mitigation/Prevention
- Preparedness
- Response
- Recovery
- Evaluation



# Humanitarian logistics

- Process of planning, implementing and controlling the efficient, cost-effective flow of and storage of goods and materials as well as related information, from the point of origin to the point of consumption, for the purpose of alleviating the suffering of vulnerable people



# Humanitarian logistics

	Assessment & planning	Procurement	Warehousing	Transportation
Strategic decisions (>2years)	Strategic planning	Long term agreements Standard definitions	Planning capacity <b>Warehouses network (location)</b>	Planning strategy, transport capacity and network
Tactical decisions (6months -2y.)	Planning demand, activities, standard equipm	Procurement methods, <b>advanced procurement</b>	<b>Planning prepositioning &amp; management</b>	Planning modes and transportation routes
Operational decisions (<6 months)	Damage & needs assessment , local capacities assessment, launch SAR, deployment of teams	<b>Acquisition, consolidate orders, new orders</b> , defining standard and kits, agreements	Managing goods, preparing kits, control of inflows and outflows, location on the field	<b>Transport primary (point to point) and secondary movements (to several EDPs or warehouses on the field).</b> Customs

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# Preparedness: strategic and tactical decisions

- Disasters can be **seasonal** (rains, cyclones, droughts), or non seasonal (earthquakes,...)
  - Many countries suffer **recurrent natural disasters**
  - Disasters can be slow onset or **sudden onset**
  - Authorities must **provide relief aid** to affected people under a **very high time pressure: prepositioning**
- 
- The image shows a group of approximately 15 people on a small, white motorboat on a wide, muddy river. Most of the people are wearing bright orange life jackets. They appear to be engaged in a relief operation, with some individuals handling supplies and others looking towards the camera. The boat is filled with various items, including bags and what looks like a motorcycle. The background shows a wide riverbank with some sparse vegetation under a cloudy sky.
- Big amounts of **relief aid must be moved**
  - Logistics is critical to provide relief at *right time, right place, right amount, right quality, right cost, right source (the six “rights”)*
  - **First responders:** population, civil protection, governments, NGO on the country
  - **Local capacities** critical to response (72 hours). International support will arrive with **delay**.

# Preparedness model: local entities

- **Strategic decisions** (unique for different scenarios):
  - When and where to **locate** the main warehouses?
  - What does their **capacities** have to?
- **Tactical decisions** (unique for different scenarios):
  - How much **humanitarian aid** to be prepositioned?
  - Where to **locate** the prepositioned aid?
  - What **budget** should be saved for facilities and for response?
- **Operational decisions** (for each scenario):
  - How **transport** the goods to be distributed?
  - How much humanitarian aid to be **acquired in emergency**?
  - How much humanitarian aid to be **acquired to restore stock levels**?



# Preparedness model: elements

Model for **strategical and tactical decisions** taking into account **operational decisions** under different **scenarios**

- Time horizon: several years
- Stages: each year (e)
- Periods: each season into a year (t)
- Transport network:
  - Nodes (demand, supply, transit...)
  - Links (roads)
- **Scenario**: for each season
  - **People and affected area** : **demand in each node**
  - **Roads conditions** for distributing relief aid (links capacities)

# Model description: criteria measures

## Unmet demand:

- Stochastic (for each scenario and commodity)
- Expected value per season

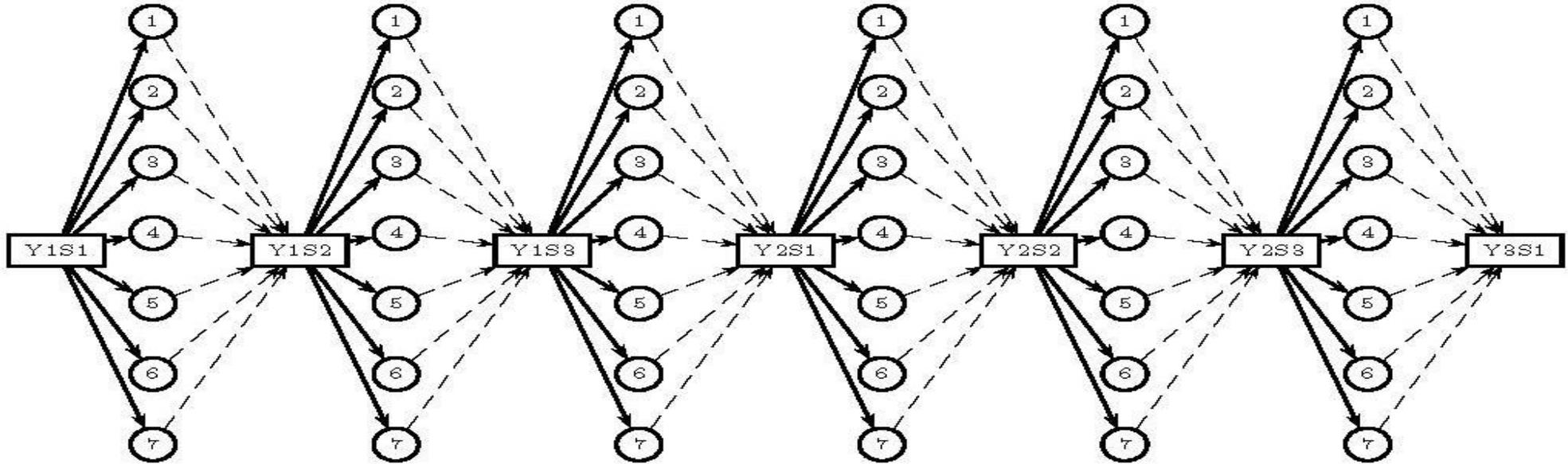
## Cost:

- Deterministic cost (strategical) + Stochastic cost (response and prepositioning)
- Expected value of budget deviations

**Response time:** Hard constraints limiting origin of aid

**MULTIOBJECTIVE OPTIMISATION: Pareto Frontier**

# Model description: scenarios



For each period, scenarios  $\Omega^t \subseteq \Omega$  ( $t \in T$ ). For each  $\omega \in \Omega^t$

- a weight or probability  $W^\omega$ ,
- demand of commodity  $o$  in the network node  $i$ ,  $D^{i\omega}_o$
- capacity of network link  $a$ ,  $N^\omega_a$ , and cost  $C^\omega_{o,a}$

# Preparedness model: variables

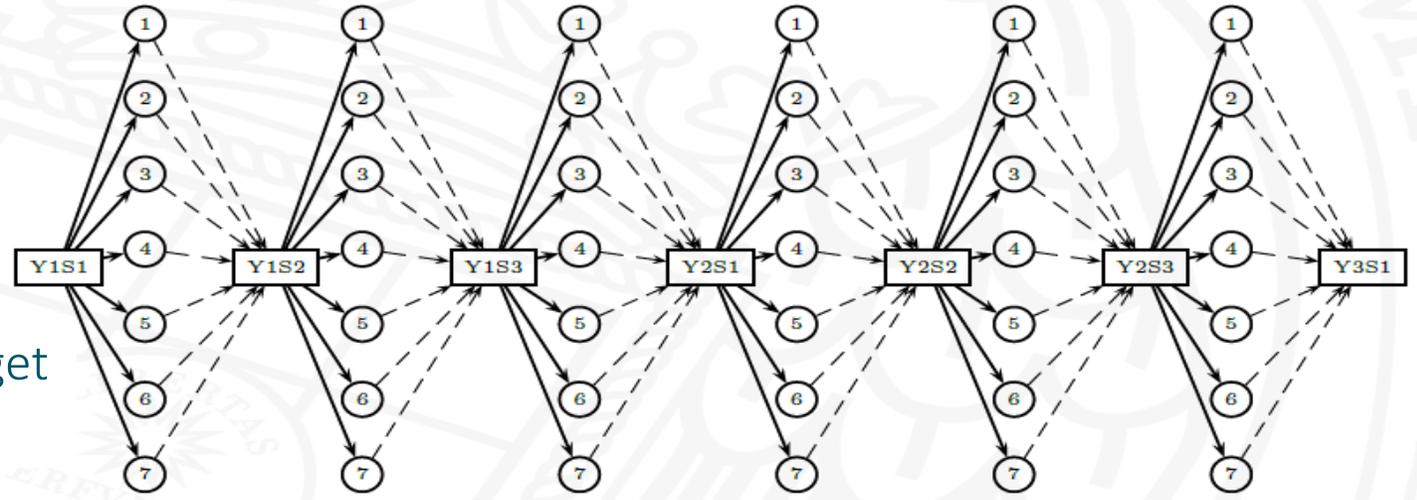
$\alpha_i^e$  warehouses location

$f_i^e$  warehouses capacities

$\beta_F^e$  facilities budget

$\beta_T^t$  procurement and distribution budget

$\hat{s}_{oi}^t$  prepositioning



$$\gamma_0 s_{oi}^\omega + r_{oi}^\omega - \tilde{r}_{oi}^\omega = \hat{s}_{oi}^{t+1}$$

$$\begin{aligned} & \hat{s}_{oi}^t + x_{oi}^\omega + \sum_{a \in \Gamma_i^{-\omega}} y_{oa}^\omega = \\ & = D_{oi}^\omega - \boxed{v_{oi}^\omega} + \sum_{a \in \Gamma_i^{+\omega}} y_{oa}^\omega + s_{oi}^\omega \quad \forall o \in \mathcal{O}, i \in \mathcal{I}, \omega \in \Omega^t, t \in \mathcal{T} \end{aligned}$$

# Preparedness model: Budget allocation

$$\bar{C}^e \geq \beta_F^e + \sum_{t \in \mathcal{T}^e} \beta_T^t \quad \forall e \in \mathcal{E}$$

$$\sum_{i \in \mathcal{I}} B_i^e (\alpha_i^e - \alpha_i^{e-1}) + F_i^e (f_i^e - f_i^{e-1}) + \sum_{i \in \mathcal{I}} \tau_i (B_i^e \alpha_i^e + F_i^e f_i^e) = \beta_F^e \quad \forall e \in \mathcal{E}$$

$$\sum_{o \in \mathcal{O}} \sum_{i \in \mathcal{I}} (Q_{oi}^t (r_{oi}^\omega - \tilde{r}_{oi}^\omega) + C_{oi}^t x_{oi}^\omega) + \sum_{o \in \mathcal{O}} \sum_{a \in \mathcal{A}^\omega} G_o^L C_{oa}^\omega y_{oa}^\omega \leq \beta_T^t + v_T^\omega \quad \forall \omega \in \Omega^t, t \in \mathcal{T}$$

# Preparedness model: multiobjective programming- weighting method

$$\min \frac{\lambda}{Z_C - Z^C} \cdot \left( \sum_{t \in \mathcal{T}} \frac{1}{(1 + \rho)^t} \sum_{\omega \in \Omega^t} W_\omega v_T^\omega - Z_C \right) +$$
$$\frac{1 - \lambda}{Z_D - Z^D} \cdot \left( \sum_{o \in \mathcal{O}} M_o \sum_{\substack{t \in \mathcal{T}, \\ \omega \in \Omega^t}} \sum_{i \in \mathcal{I}} W_\omega v_{oi}^\omega - Z_D \right)$$

# Preparedness model: scenarios generation

## Scenario features:

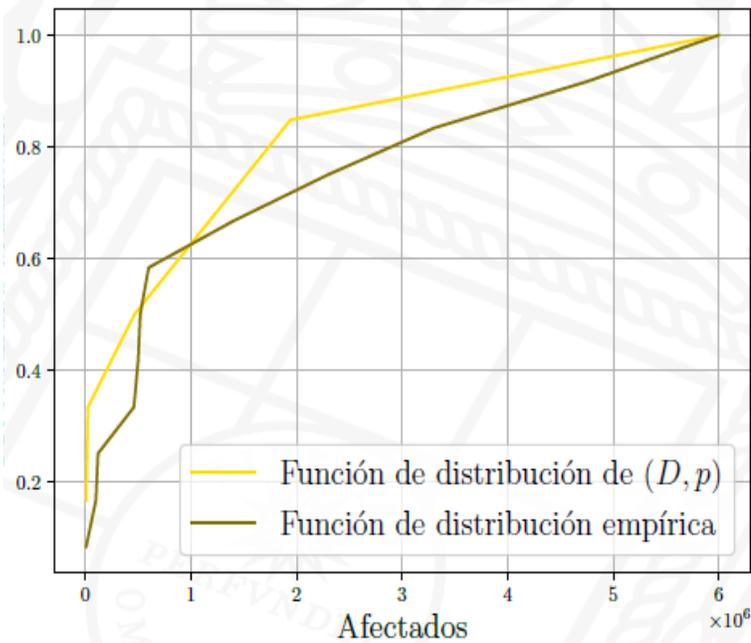
- a weight or **probability**,
- **demand** of commodity  $o$  in the network node  $i$
- **capacity** of network link  $a$  and transport cost of commodity  $o$  through link  $a$

## Characteristics:

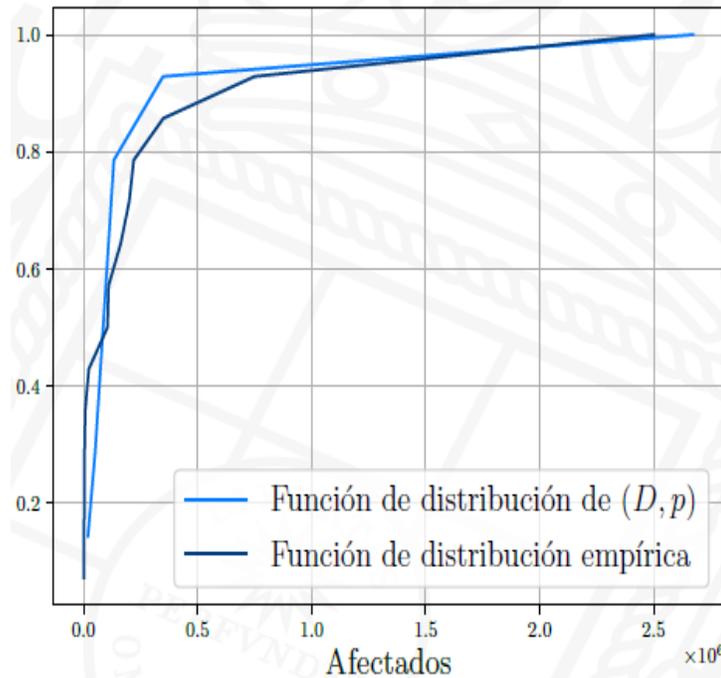
- Recurrent disasters: **Historical data**
- **Disaster type data**: consequences very dependent on the type of disaster
- **Geographical data**: disasters and their consequences very dependent on the location
- **Scarce reliable information** (Unlike what happens today in data science)
  - Especially developing countries: bad or few information registered
- **Several data sources**

# Scenarios generation: Phase 1- Global demand classes

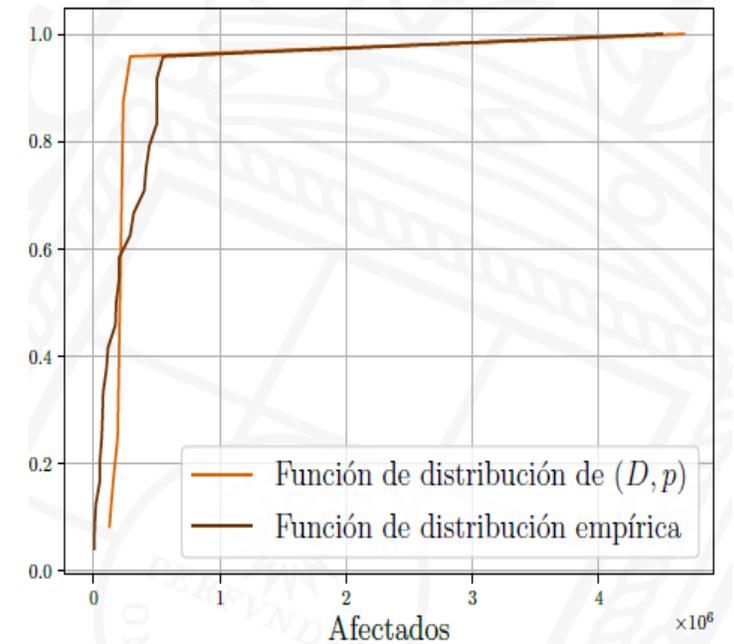
*Drought*



*Storm*



*Flood*



# Flood scenarios

Data
300
500
3500
6925
17000
47837
50000
63946
70000
75000
100003
113535
171600
177645
200000
200000
288500
315986
400000
415000
440000
500000
500000
500000
549326
4500000

Ph1 →

s1	5281,8
s2	17949,6
s3	61000,1
s4	207303,6
s5	704503,2
s6	4604972,2

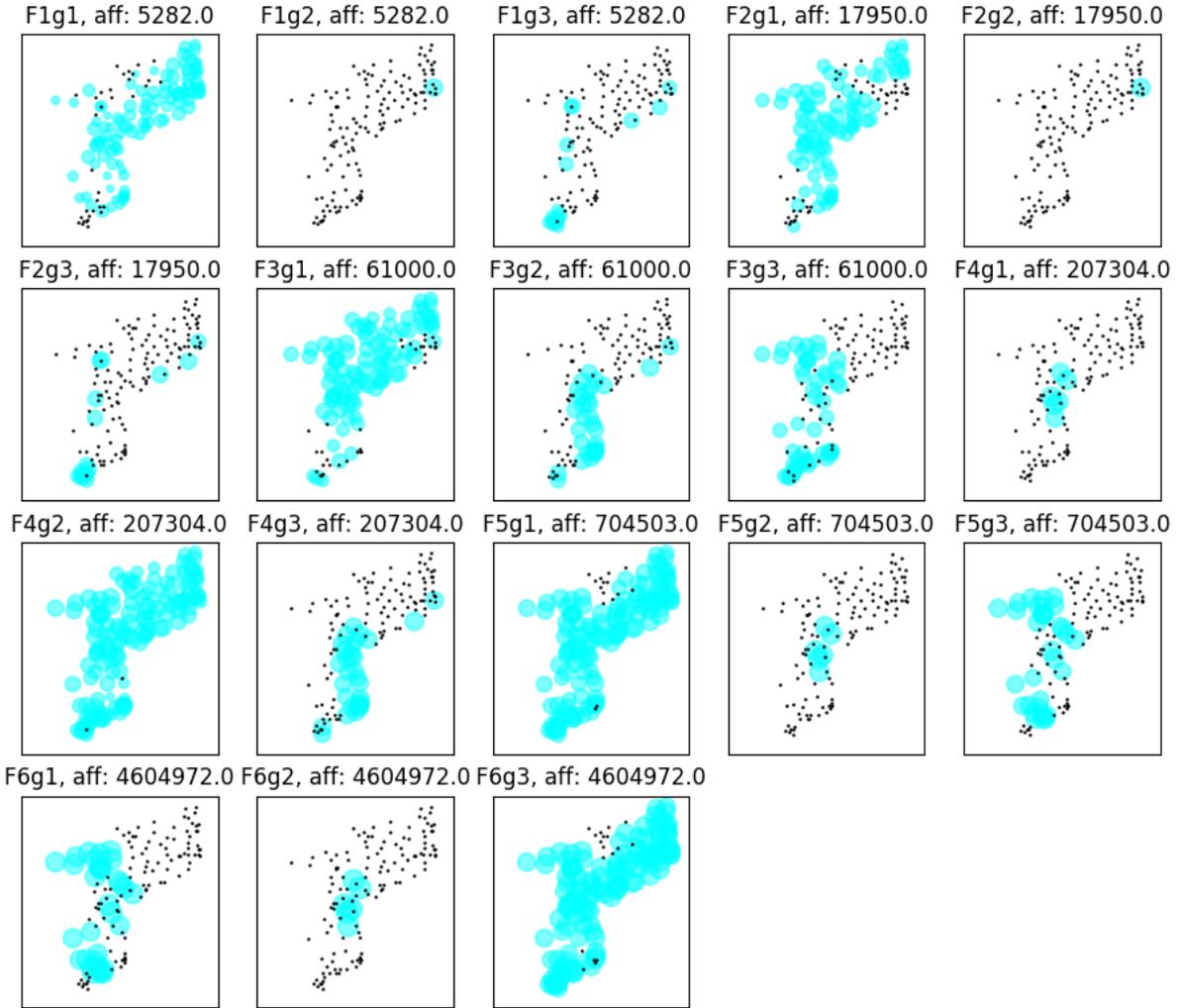
Ph2 →

s1 - 5282	s2 - 17950	s3 - 61000	s4 - 207304	s5 - 704503	s6 - 4604972
0,090977	0,222277	0,419718	0,638818	0,818243	0,958249
0,086851	0,218150	0,415592	0,634691	0,814117	0,954123
0,028083	0,159383	0,356825	0,575924	0,755350	0,895355
0,022696	0,108603	0,306045	0,525144	0,704570	0,844576
0,123896	0,007404	0,204845	0,423945	0,603370	0,743376
0,285821	0,154521	0,042920	0,262020	0,441445	0,581451
0,293538	0,162238	0,035203	0,254303	0,433728	0,573734
0,337191	0,205892	0,008450	0,210649	0,390075	0,530080
0,353482	0,222182	0,024741	0,194359	0,373784	0,513790
0,365967	0,234667	0,037226	0,181874	0,361299	0,501305
0,418301	0,287002	0,089560	0,129539	0,308965	0,448970
0,441365	0,310065	0,112624	0,106476	0,285902	0,425907
0,515186	0,383887	0,186445	0,032654	0,212080	0,352086
0,521234	0,389934	0,192493	0,026607	0,206032	0,346038
0,541716	0,410417	0,212975	0,006124	0,185550	0,325556
0,541716	0,410417	0,212975	0,006124	0,185550	0,325556
0,602375	0,471076	0,273634	0,054535	0,124891	0,264896
0,616709	0,485409	0,287968	0,068868	0,110557	0,250563
0,652272	0,520973	0,323531	0,104432	0,074994	0,214999
0,657609	0,526309	0,328868	0,109768	0,069657	0,209663
0,665962	0,534663	0,337221	0,118122	0,061304	0,201309
0,683668	0,552368	0,354926	0,135827	0,043599	0,183604
0,683668	0,552368	0,354926	0,135827	0,043599	0,183604
0,683668	0,552368	0,354926	0,135827	0,043599	0,183604
0,696203	0,564904	0,367462	0,148363	0,031063	0,171068
0,866406	0,735107	0,537665	0,318566	0,139140	0,000865

# Flood scenarios

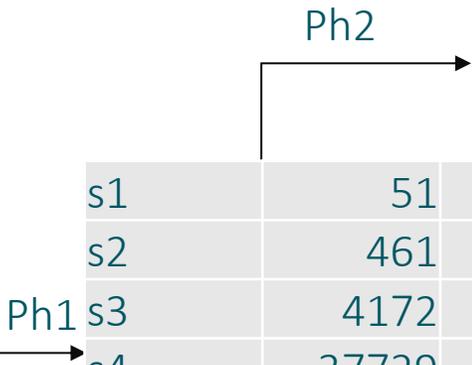
Data
300
500
3500
6925
17000
47837
50000
63946
70000
75000
100003
113535
171600
177645
200000
200000
288500
315986
400000
415000
440000
500000
500000
500000
549326
4500000

Ph3 →

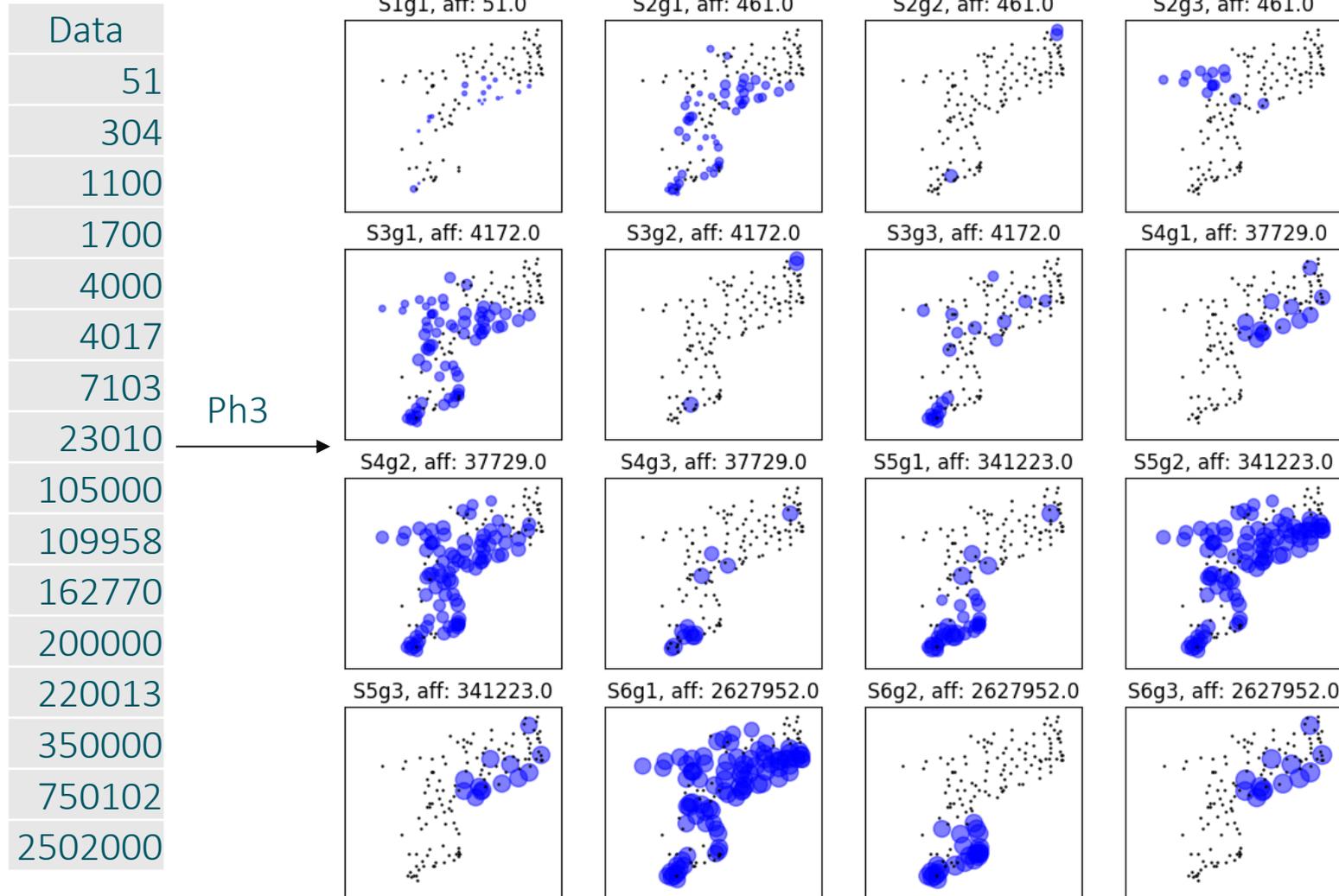


# Storm scenarios

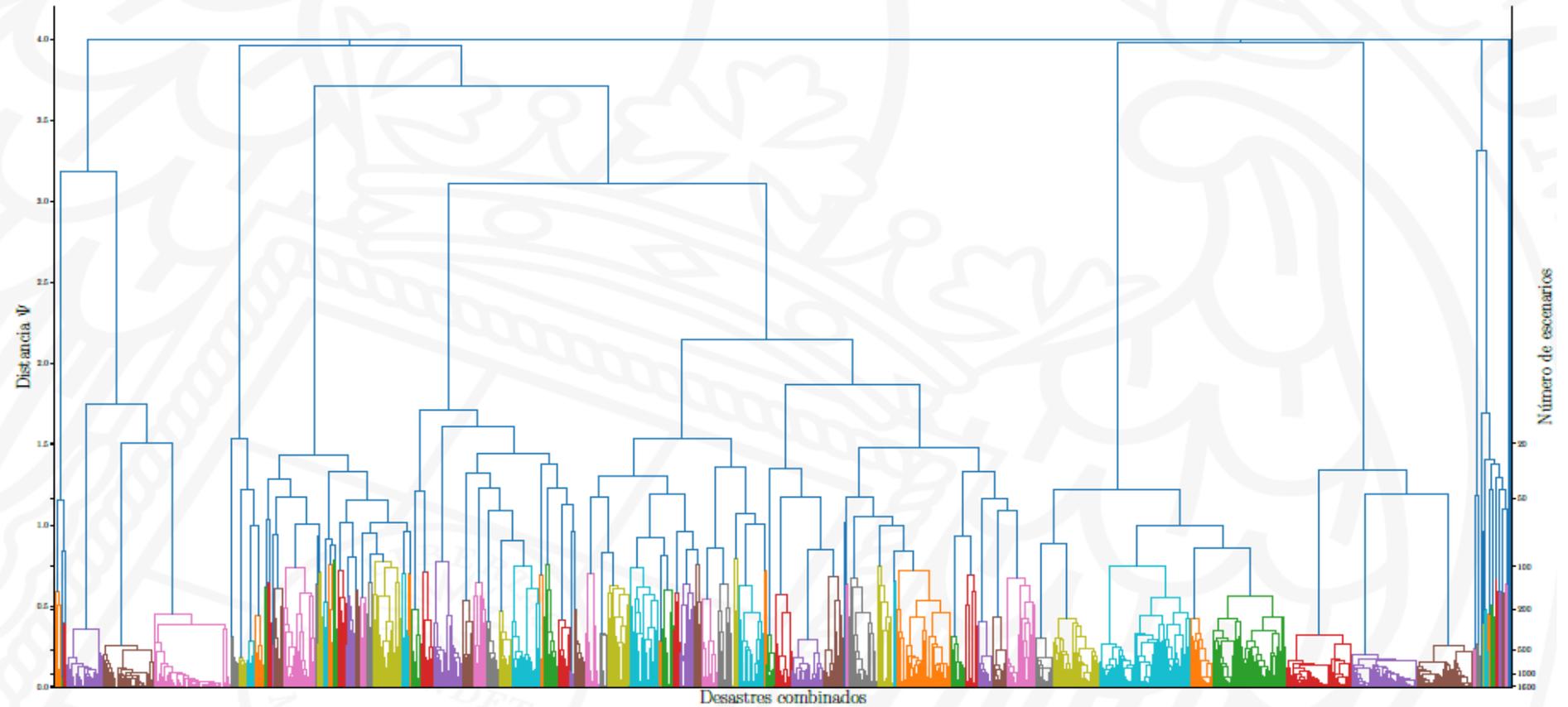
Data		s1 - 51	s2 - 461	s3 - 4172	s4 - 37729	s5 - 341223	s6 - 2627952
51		0	0,078347857	0,267597586	0,544686362	0,790991051	0,917898389
304		0,056209402	0,022138455	0,211388184	0,48847696	0,734781649	0,861688987
1100		0,138387993	0,060040136	0,129209593	0,40629837	0,652603059	0,779510396
1700		0,175768058	0,097420202	0,091829527	0,368918304	0,615222993	0,74213033
4000	s1 51	0,262885332	0,184537475	0,004712254	0,28180103	0,528105719	0,655013057
4017	s2 461	0,263358363	0,185010507	0,004239222	0,281327999	0,527632688	0,654540025
7103	s3 4172	0,330033588	0,251685732	0,062436003	0,214652774	0,460957463	0,5878648
23010	s4 37729	0,480318172	0,401970315	0,212720586	0,064368191	0,31067288	0,437580217
105000	s5 341223	0,670627478	0,592279621	0,403029892	0,125941116	0,120363573	0,247270911
109958	s6 2627952	0,675917808	0,597569951	0,408320222	0,131231445	0,115073244	0,241980581
162770		0,719131848	0,640783992	0,451534263	0,174445486	0,071859203	0,198766541
200000		0,74046318	0,662115323	0,472865594	0,195776818	0,050527871	0,177435209
220013		0,750000782	0,671652926	0,482403197	0,20531442	0,040990269	0,167897606
350000		0,793213755	0,714865898	0,525616169	0,248527393	0,002222704	0,124684634
750102		0,852145532	0,773797675	0,584547946	0,307459169	0,06115448	0,065752857
2502000		0,915977276	0,837629419	0,64837969	0,371290914	0,124986225	0,001921113



# Storm scenarios



# Scenarios: merge and reduction



$|\Omega| = 91$  (umbral  $\Phi = 0,8$ )

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# Case study: Mozambique

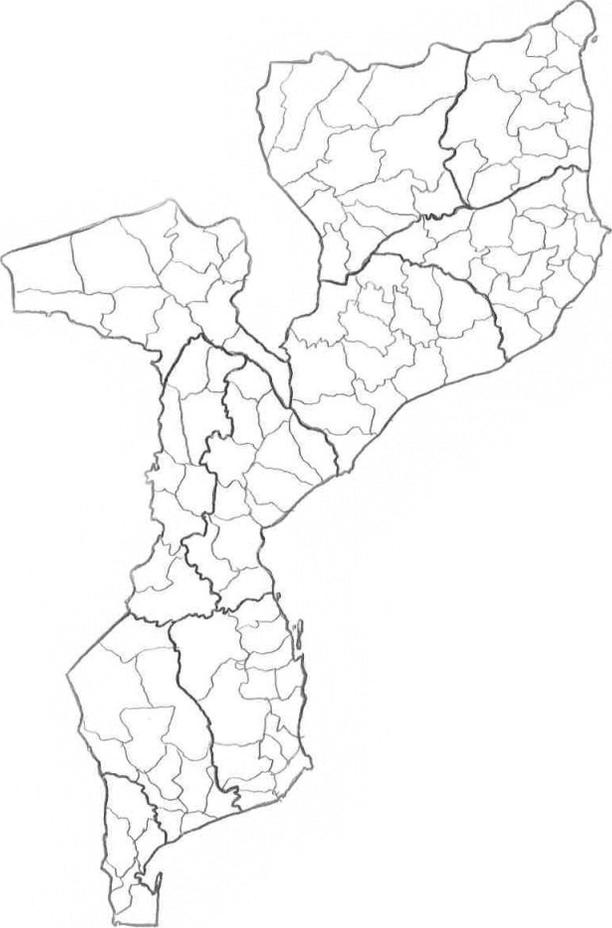
- Data sources:
  - EMDAT
  - Desinventar
  - Google maps
  - Mozambican entities (meteorology, transport...)
  - Logistic cluster
  - NGOs, Red Cross...



# Case study: Mozambique

- Disasters: **Flood, Storm, Drought**, Epidemic, Earthquake
- Slow onset(no roads damages): Drought, Epidemic
- Rapid onset: **Flood, Storm**, Earthquake (no data)
  
- Databases: EMDAT, DESINVENTAR
- Few information, high contradiction and not accurate information
- Big efforts to **cross information from both databases**

# Case study: Mozambique



# Case study: Mozambique

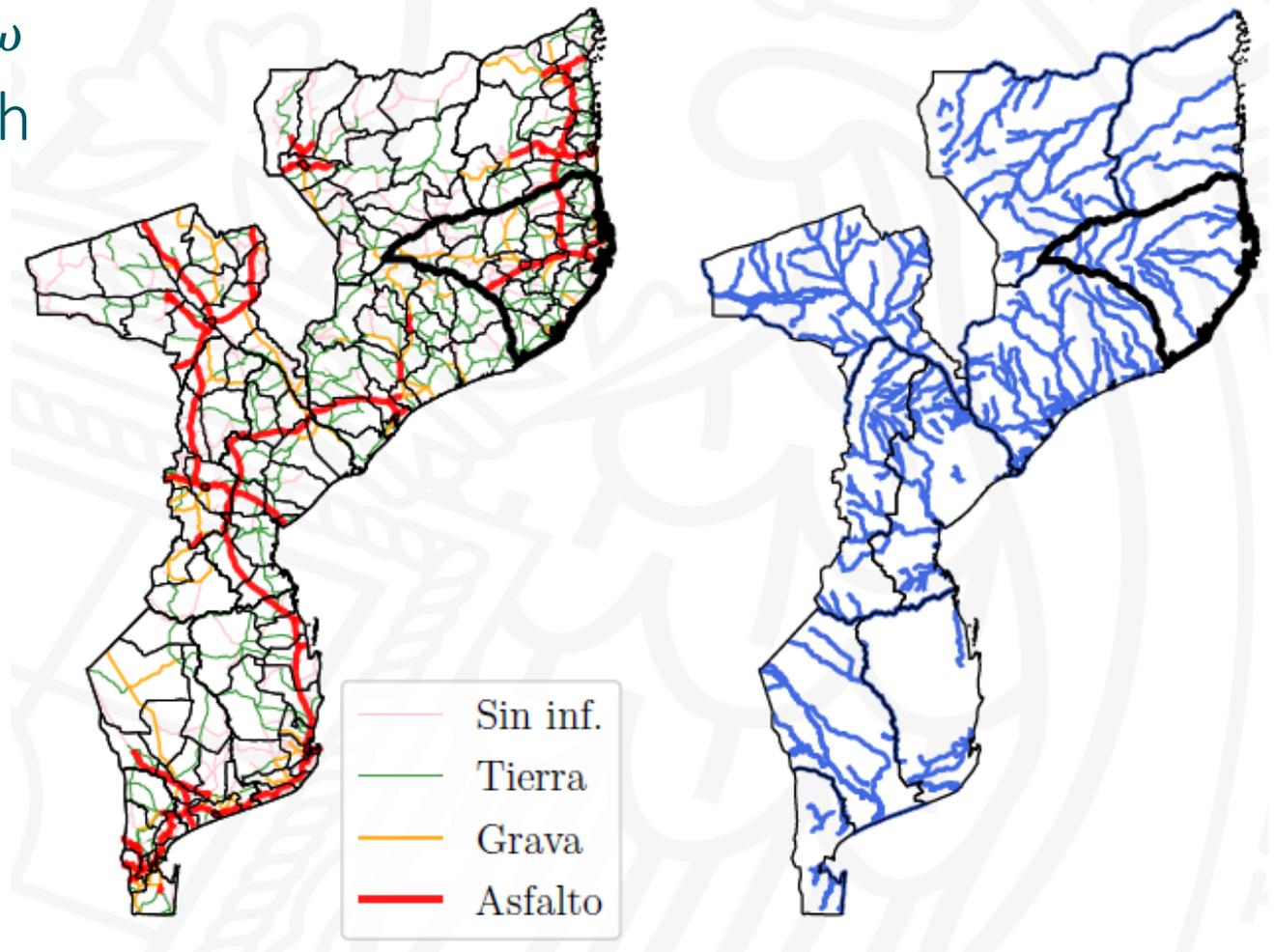


Nodes: 128 districts

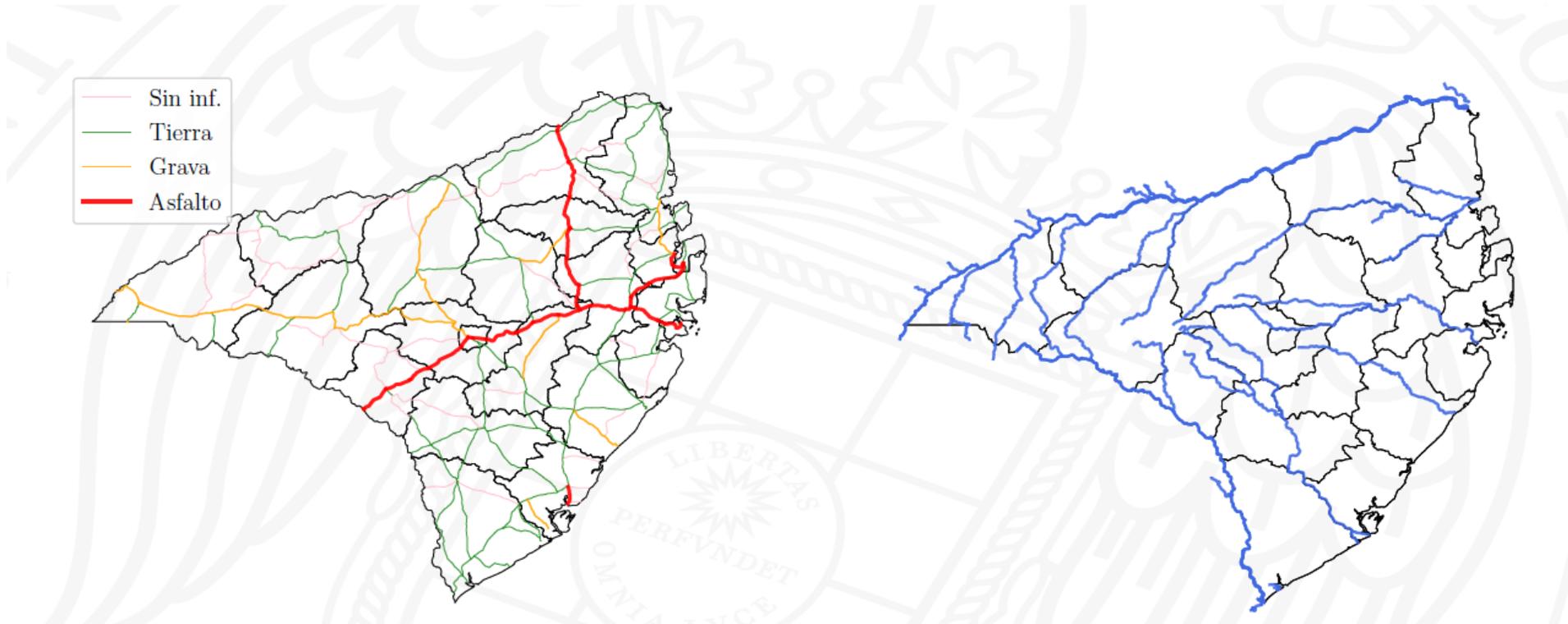
Links: 278

# Case study: Mozambique

- Capacity of network link  $a$ ,  $N_{a,\omega}$
- Unit transportation cost through link  $a$ ,  $C^0_{a,\omega}$



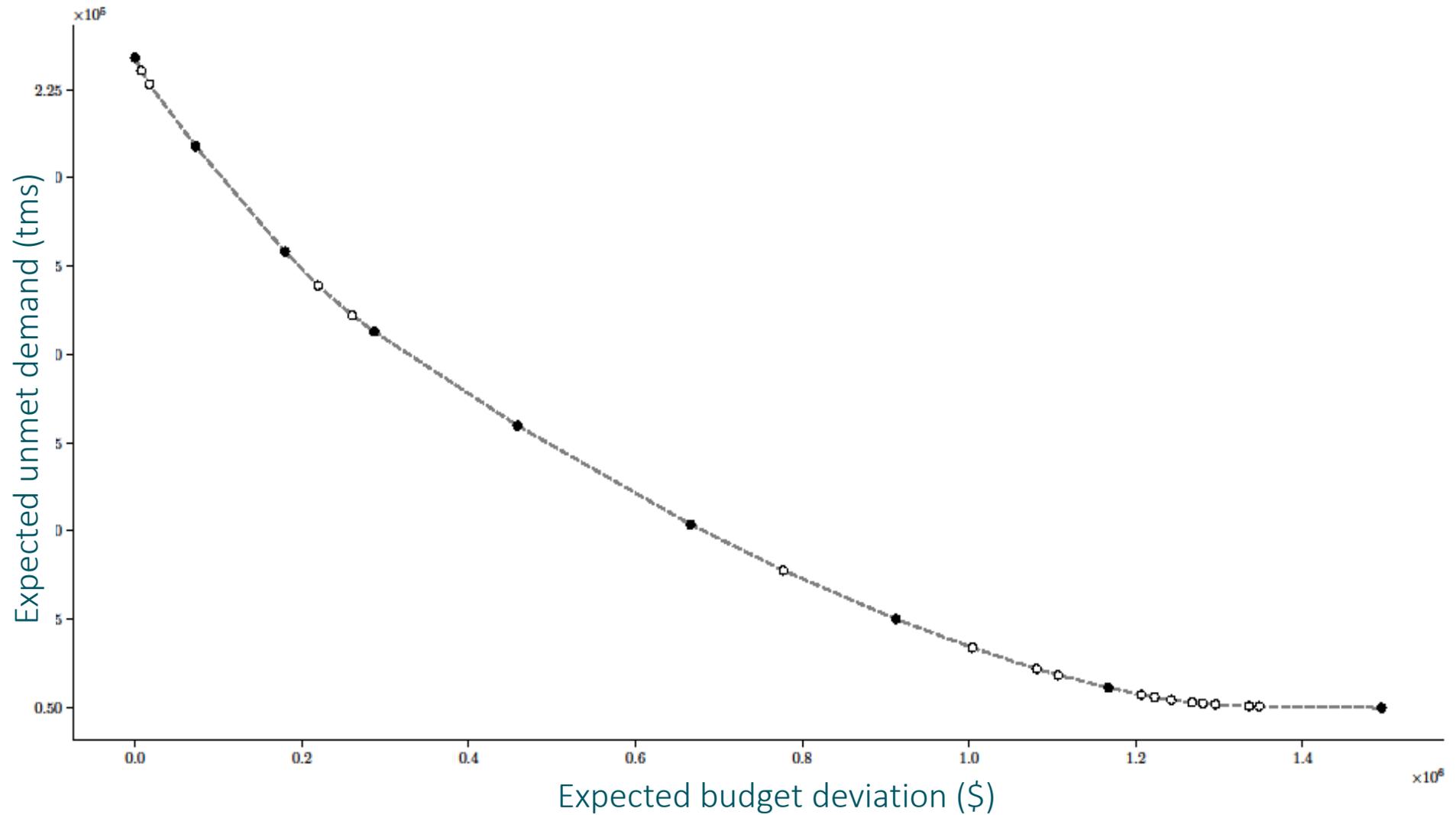
# Case study: Nampula



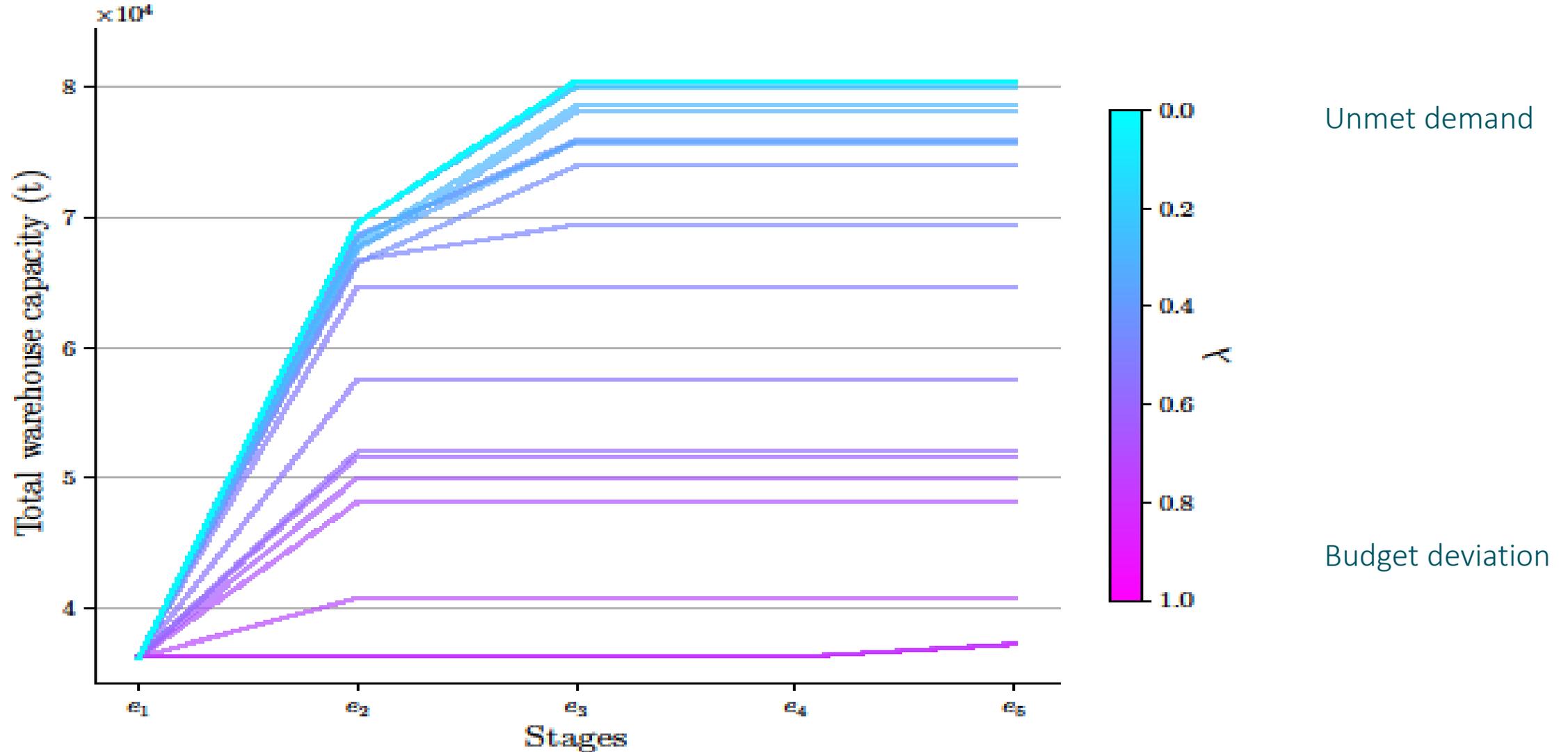
# Case study: Nampula

- 21 nodes y 78 links.
- 35 scenarios (19 in June-July-August).
- 5 years with 4 seasons per year.
- 183 733 rows, 259 288 columns, and 105 binary variables.

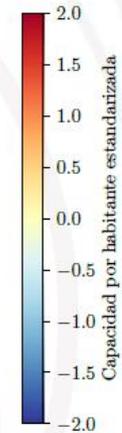
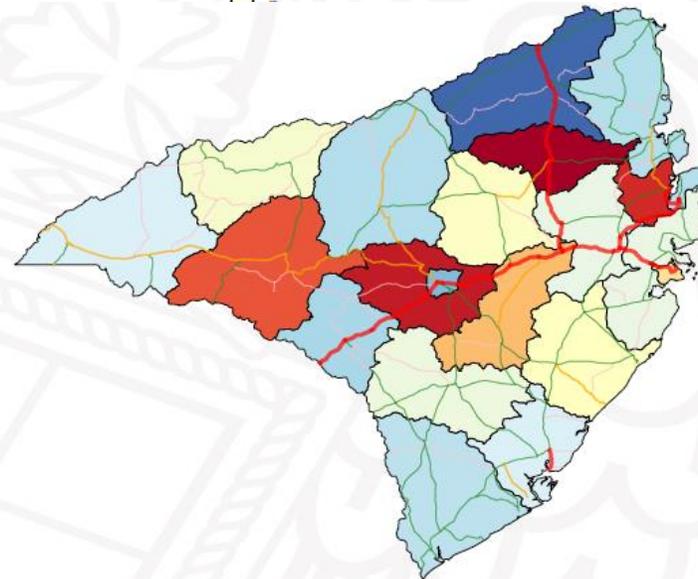
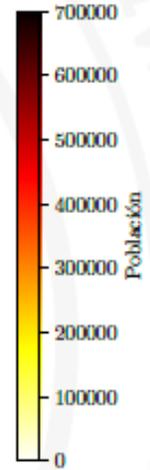
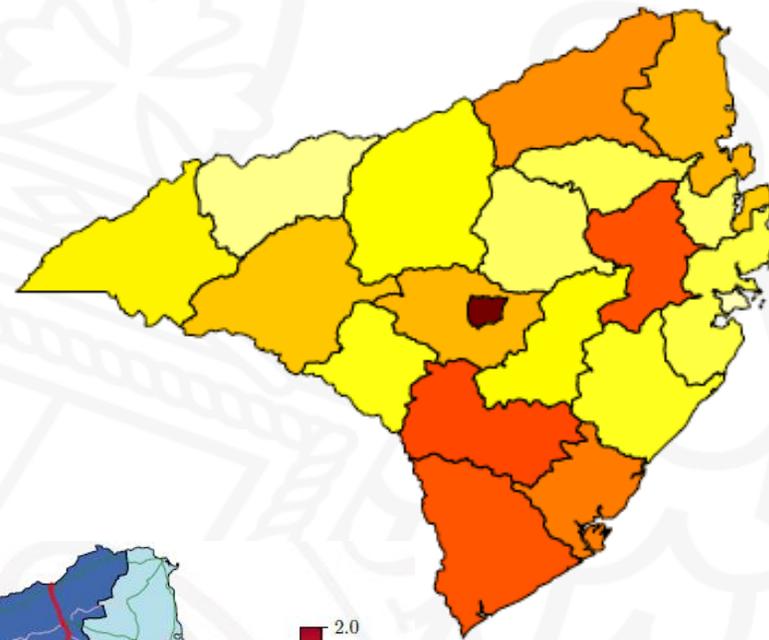
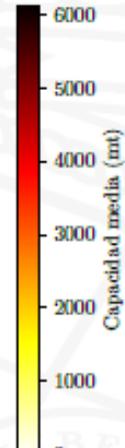
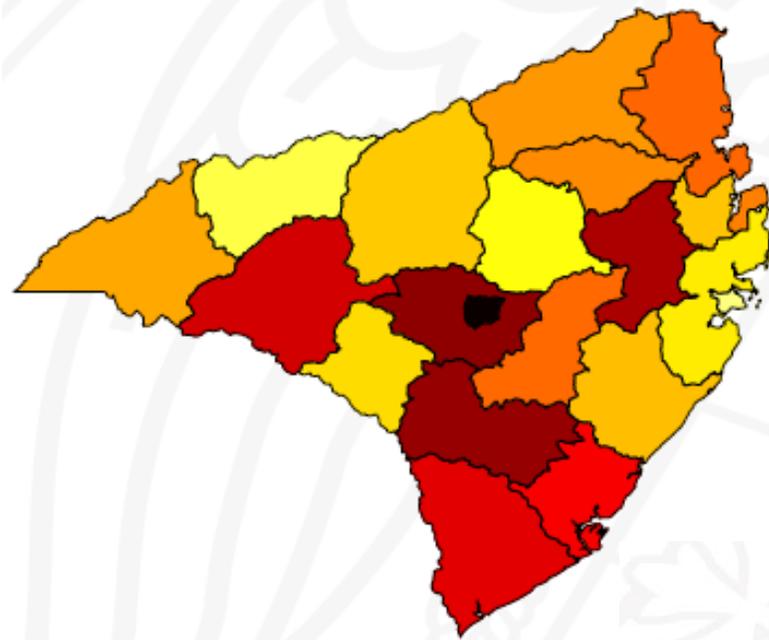
# Pareto Frontier



# Warehouse capacities in Pareto Frontier



# Warehouse capacities installed



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# Exploring alternatives

- Objective:
  - Solving the model for full Mozambique
  - Other measures (CVaR...)
  - Working with INGD of Mozambique (research project, students...)
- Difficulties: high running times
- Alternatives for modelling and methodologies:
  - Breaking symmetries: Different building costs (the further away from the airport/port the more expensive it is)
  - Weighting method vs e-constraints: e-constraints looks better
  - Impulse variables vs pulse variables
- Alternatives for solving:
  - Benders decomposition
  - Heuristics

# Experiments

- Experiment 1: comparing impulse-pulse variables with runtime limit 20 hours

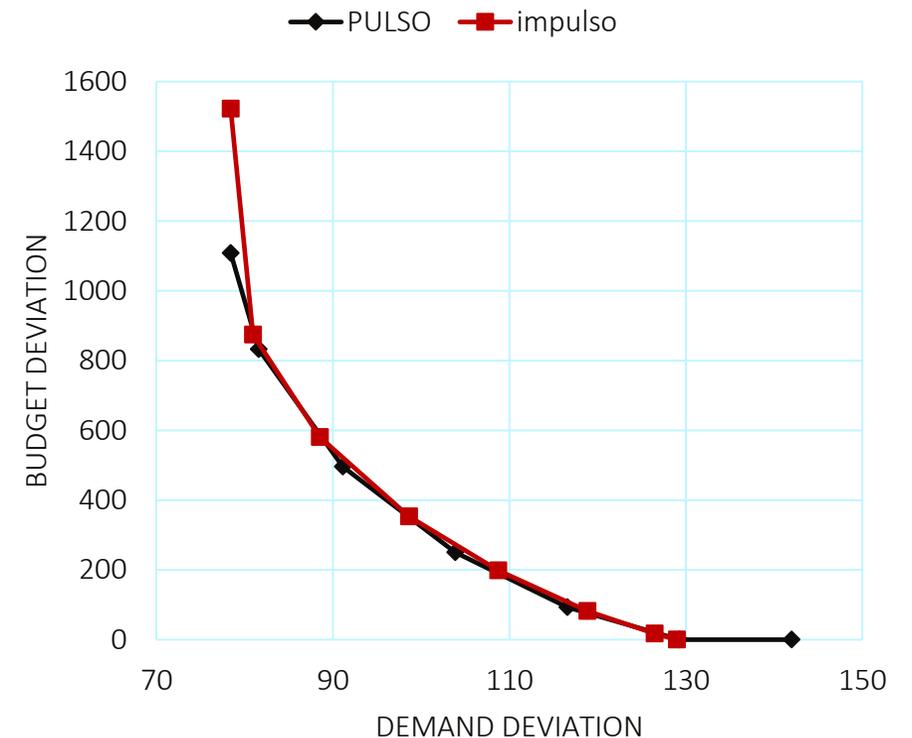
- Minimising cost, e-constraint demand
- Findings:
  - Payoff matrix pulse-payoff matrix impulse

78,425	1108,293	78,425	1522
142,001	0	128,979	0

- Similar Pareto Frontier
- Incumbent solutions: no significant differences after 2h

- Conclusion:

- Comparison limiting runtime 2 hour
- Affect values (initial solution in loops different)
- Breaking symmetries

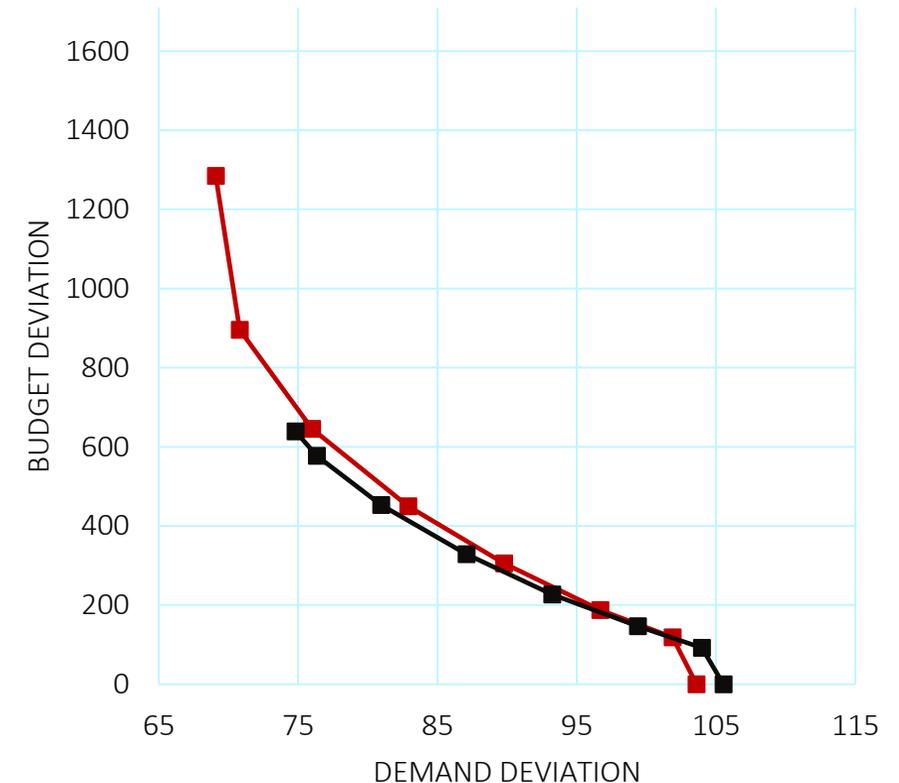


# Experiments

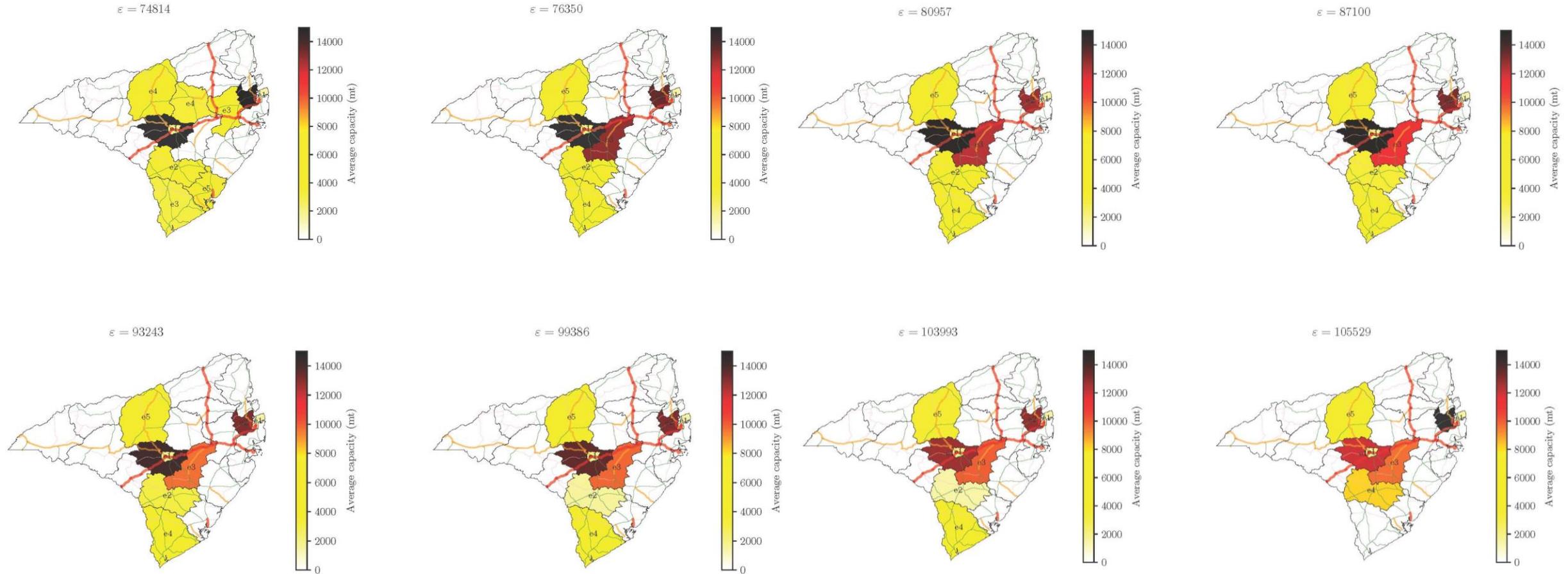
- Experiment 2: comparing impulse-pulse variables with runtime limit 1 hour
  - Building cost changing with distance to airport/port
  - Minimising cost, e-constraint demand
  - Findings:
    - Payoff matrix pulse-payoff matrix impulse

74,81	639,11	69,09	1284,61
105,53	0	103,6	0

- Including lower bound for demand can be worse (639,11)
- Difficult to change binary variables in loop for Pareto Frontier



# Solution with pulse variables and asymmetric costs



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# Conclusions

- Humanitarian logistics is key to provide relief aid to people in need
- A multicriteria model has been developed for:
  - Strategic decisions: relief aid warehouses location and sizing
  - Tactical decisions: relief aid prepositioning and budget reserved
- considering:
  - Operational decisions on different scenarios
  - Expected unmet demand (72h)
  - Expected budget deviation (72h)
- Runtimes are too high
- Work made reducing times: pulse formulation, e-constraints, changes in costs
- Work in progress: Benders decomposition
- Future objective: solving the full case Mozambique, validate it with the INGD

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