

# Insurance, Probabilities & Natural Catastrophes

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# Insurance and Uncertainty : the Framework

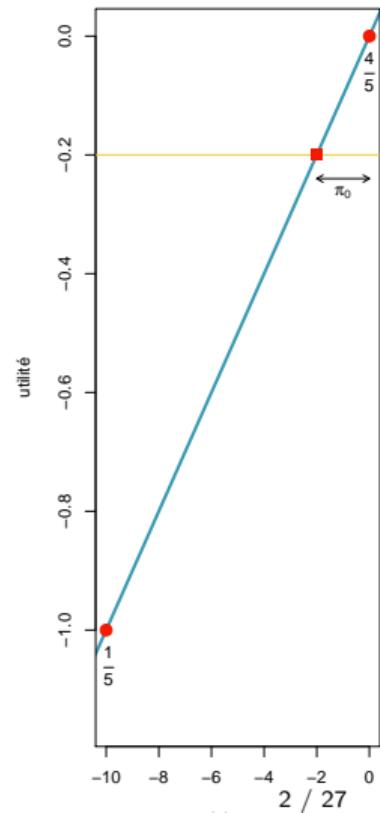
Classical framework : agent with no risk aversion facing random loss  $L$ , purchases insurance at price  $\pi_0$  if

$$\pi_0 = \mathbb{E}_{\mathbb{P}}(L) = p \ell$$

if a loss  $\ell$  can occur with probability  $p$ .

$$\{(\ell, p), (0, 1 - p)\} \sim \{(\pi_0, 1)\}$$

$\pi_0 = p \ell$  is the **Actuarial (pure) Premium**, also called **Learned Hand formula** in law business (see [Grossman et al. \(2006\)](#)) or simply “*probability times consequence*” in climate literature (see [Schneider \(2002\)](#))



# Insurance and Uncertainty : the Framework

Classical framework : agent, with utility  $u(\cdot)$  facing random loss  $L$ , purchases insurance at price  $\pi$  if

$$\mathbb{E}_{\mathbb{P}}[u(-\pi)] \geq \mathbb{E}_{\mathbb{P}}[u(-L)]$$

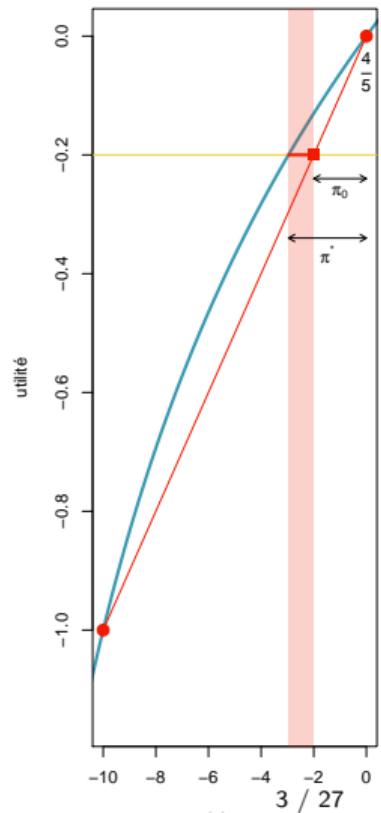
(**von-Neumann & Morgenstern** expected utility).

Let  $\pi^*$  denote the highest premium

$$\pi^* = -u^{-1}(\mathbb{E}_{\mathbb{P}}[u(-L)]) = -u^{-1}(p u(-\ell))$$

$$\{(\ell, p), (0, 1-p)\} \sim \{(\pi^*, 1)\}$$

with  $\pi^* \leq \pi_0$



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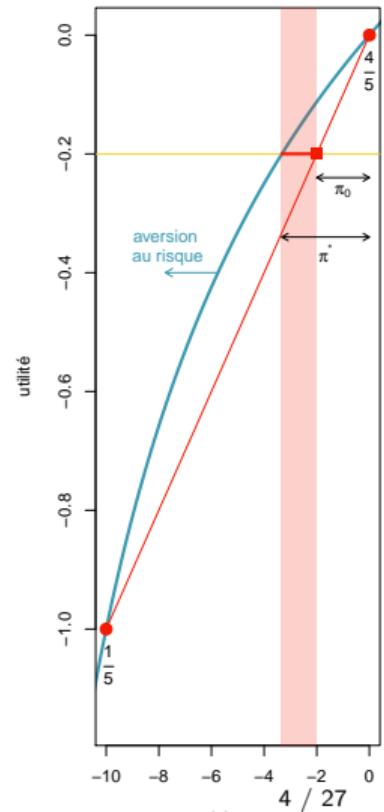
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## Insurance and Uncertainty : the Framework

Consider the simplistic case of homogeneous agents.

An **agent**, with utility  $u$ , will purchase insurance if

$$u(-\pi) > \textcolor{red}{p} u(-\ell)$$

Assume that this inequality is satisfied.

An **insurance company**, with utility  $v$ , will sell insurance if

$$\mathbb{E}_{\textcolor{red}{p}}[v(\kappa + n\pi - S)] > v(\kappa), \quad S = \sum_{i=1}^n L_i = N_n \cdot \ell, \quad \text{where } N_n \sim \mathcal{B}(n, \textcolor{red}{p}),$$

where  $\kappa$  is the capital of the company, and  $S$  is the total indemnity.

The actuarial fair premium is obtained when  $v$  is linear :

$$\pi^* = \mathbb{E}[I(L)] = \textcolor{red}{p} \cdot I(\ell)$$

# Insurance and Uncertainty : the Framework

If risks are exchangeable  $X = L_1 + \cdots + L_n = N_n \cdot I(\ell)$ ,

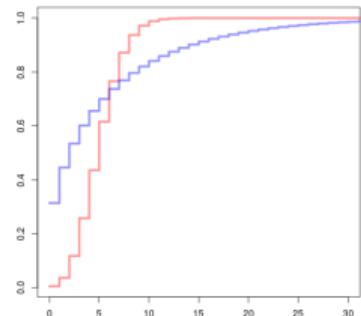
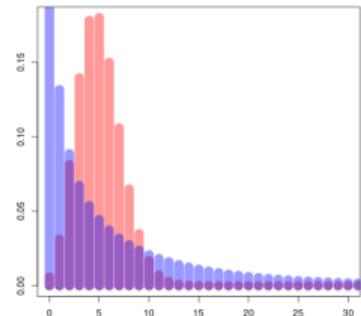
$$\mathbb{P}[N_n = k] = \int_0^1 \binom{n}{k} \theta^k (1-\theta)^{n-k} dG(\theta), \quad \mathbb{P}[L_i > 0] = p$$

from [de Finetti \(1921\)](#). See [Charpentier & le Maux \(2014\)](#) for correlated risks and disaster (with endogeneous default probability of the insurer).

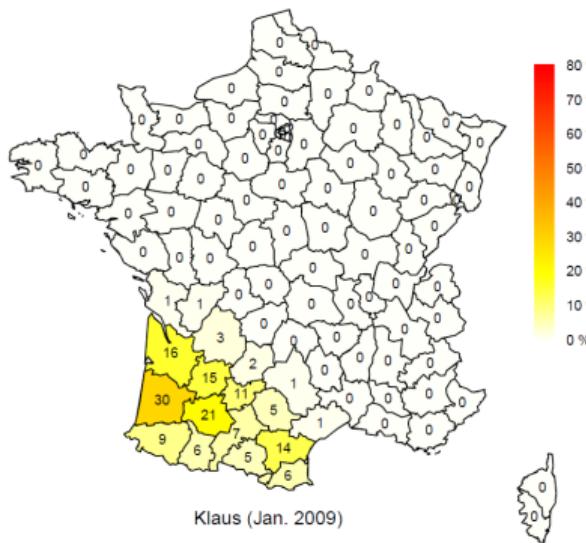
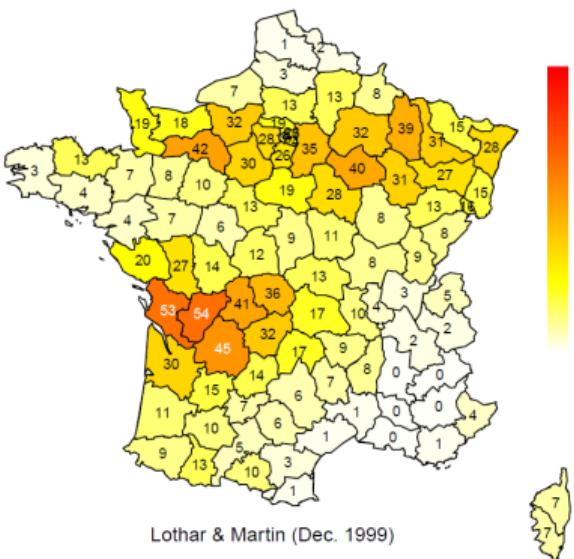
Classical framework : [binomial-beta](#) model ( $G$  is a Beta distribution  $\mathcal{B}(\alpha, \beta)$ ), then  $r = \text{corr}(X_i, X_j) = (1 + \alpha + \beta)^{-1}$ . Then

$$\mathbb{E}[N_n] = np \text{ while } \text{Var}[N_n] = (n + n(n-1)r)p(1-p)$$

(the insurance company has correlation aversion, [Richard \(1975\)](#), not clear for insured...)



## Insurance and Uncertainty : the Framework



Proportion of insurance policy that did claim a loss after storms, for a large insurance company in France (~1.2 million household policies)

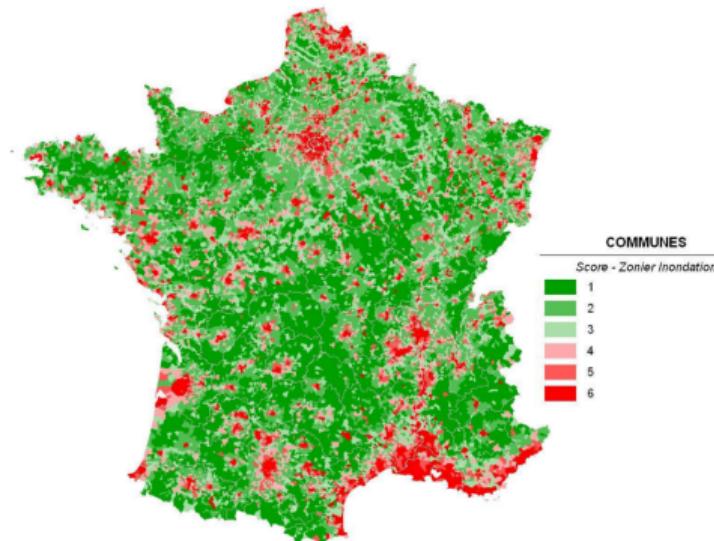
# Insurance and Uncertainty : the Framework

Consider the more complex case of heterogeneous agents, with **observed heterogeneity**

	A	B	C	D	E
probability	$p_A$	$p_B$	$p_C$	$p_D$	$p_E$
pure premium	$p_A \ell$	$p_B \ell$	$p_C \ell$	$p_D \ell$	$p_E \ell$

but, with **unobserved heterogeneity**

	A	B	C	D	E
probability	$p_A$	$p_B$	$p_C$	$p_D$	$p_E$
pure premium	$p \ell$				



where

$$p = \frac{n_A p_A + n_B p_B + n_C p_C + n_D p_D + n_E p_E}{n_A + n_B + n_C + n_D + n_E}$$

# The Challenge of Predictive Models (for Rare Events)

*"A 30% chance of rain tomorrow: How does the public understand probabilistic weather forecasts? "* Gigerenzer et al. (2005)

See Nate Silver's "there's an awful lot of room to debate what 'probably' means"

538.com, or Ronald Fisher on predicting probabilities on 'one-shot' events,

(or more provocative "*probability does not exist*" by Bruno de Finetti, Nau (2011))

Classical issue in risk management : predict 1% chance of occurrence, and occurs.

Was 1% wrong ?

Disasters are not *per se* 'one-shot' events, but we have to deal with changing environment...

Here we focus on rare events :

- (1) hard to find a good model
- (2) hard to assess if the model is good



## So what...?

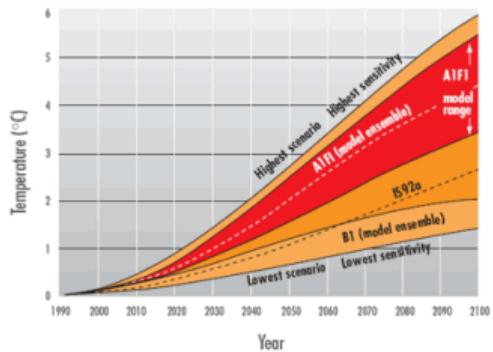
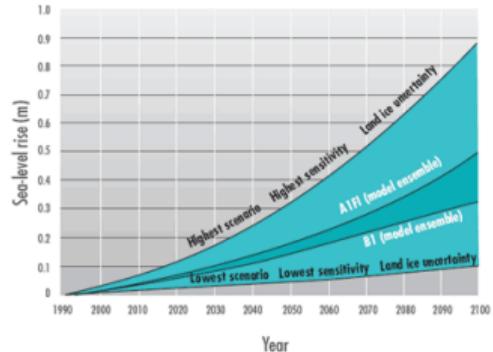
In this talk, “*natural catastrophe*” is seen from an insurance perspective

- quantitative analysis of natural events  
difficult part : climate is changing (warming)  
classical curves from **IPCC** on temperature or sea level  
(and references therein)

- impact on the insurance industry ?

In 2015, French Federation of Insurers claimed that

“*Catastrophes naturelles : la facture des assureurs pourrait doubler d'ici 2040*” (in 25 years, +3% per year,  
see <https://www.argusdelassurance.com/>)



# Insurability of Climate Risks ?

see Berliner (1985) and Charpentier (2008)

**Table 3** Insurability criteria and related requirements according to Berliner

Insurability Criteria		Requirements
<i>Actuarial</i>	(1) Randomness of loss occurrence	Independence and predictability of loss exposures
	(2) Maximum possible loss	Manageable
	(3) Average loss per event	Moderate
	(4) Loss exposure	Loss exposure must be large
	(5) Information asymmetry	Moral hazard and adverse selection not excessive
<i>Market</i>	(6) Insurance premium	Cost recovery and affordable
	(7) Cover limits	Acceptable
<i>Societal</i>	(8) Public policy	Consistent with societal value
	(9) Legal restrictions	Allow the coverage

(Table from Bienier *et al.* (2015) - for cyber risk)

climate disasters will be insured if there is an insurance market for that...

## Insurance and Uncertainty : the disaster puzzle

Earthquakes and floods cause potentially large losses, rational people should find actuarially fair insurance policies attractive, see [Kunreuther \(1996\)](#)

Individuals *underestimate* the true probability of a disaster event occurring and/or have fairly high discount rates for the benefits of uncertain future reimbursements due to losses.

[Kunreuther & Pauly \(2004\)](#) proved that even when insurance for low-probability, high-loss events is offered at favorable premiums, the search costs associated with obtaining the information on premiums and disaster probabilities may prevent from purchasing insurance.

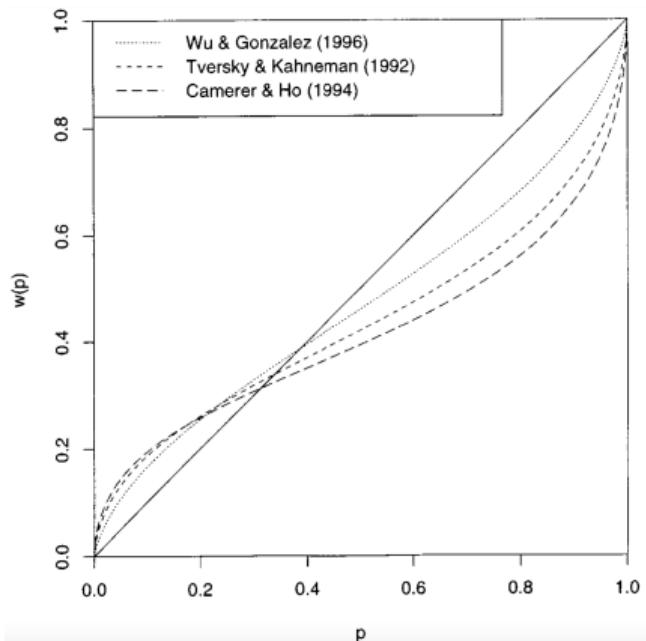
Individuals lack information about the expected harm, before the disaster, see [Botzen \(2013\)](#). Insurance prices can be a signal used to correct for possible bias (unless market distortion), see [Thieken et al. \(2006\)](#).

## Loss Probability $p$ and Perceived Loss Probability $\omega(p)$

RDEU framework, or Prospect theory, Kahneman & Tversky (1979) or Wu & Gonzalez (1996)

$$\omega(p) = \frac{p^\gamma}{(p^\gamma + (1-p)^\gamma)^{1/\gamma}}$$

(experimental) evidence of convexity of  $\omega$   
extreme risk aversion in situations that involve low stakes see Huysentruyt & Read (2010), but inadequate insurance coverage against disaster, Viscusi (2010), even if it is highly subsidized : underweighting of adverse tail events.





## Loss Probability $p$ and Perceived Loss Probability $\omega_i(p)$

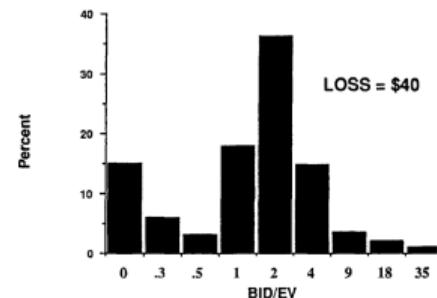
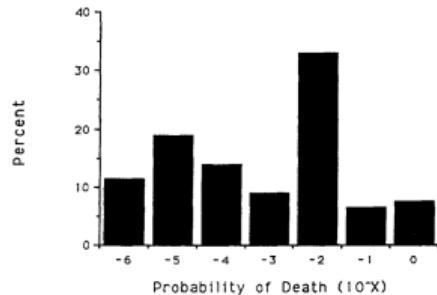
Experimental study of McClelland, Schulze & Coursey (1993)

: bimodal distribution of willingness to pay for insurance, with two groups

- neglect low-probability risks, do not purchase insurance
- willingness to pay higher than expected loss

link between the bimodal risk judgments and protective behavior McClelland, Schulze & Hurd (1990), over and underweighting rare extreme events, Epper & Fehr-Duda (2017).

See also Charpentier & le Maux (2014) on optimal (Pareto) vs. equilibrium (Nash)



## Loss Probability $p$ and Perceived Loss Probability $\omega(p)$

*"The probability of tail events is overestimated, which is consistent with probability weighting in prospect theory"* Botzen, Kunreuther & Michel-Kerjan (2015) (but "potential damage is underestimated")

Myopic loss aversion or myopic probability weighting, see Barberis, Huang & Thaler (2006), and Barberis (2013) on psychology of "tail events".

See also Gilboa & Schmeidler (1989) on the use of capacities : instead of computing  $\mathbb{E}_{\mathbb{P}}[u(\omega - \pi - L + I(L))]$ , consider

$$\min_{Q \in \mathcal{P}} \{\mathbb{E}_Q[u(\omega - \pi - L + I(L))]\} \text{ for some set of beliefs } \mathcal{P}$$

Can capture uncertainty associated with climate change, that makes it difficult to respond optimally, see Dessai & Hulme (2006) or Tompkins & Adger (2005).

## Loss Probability $p$ : Meteorological Perspective

For hurricanes, see [Gray et al. \(1992\)](#), the seasonal number of intense hurricanes is

$$\hat{N} = 3.571 + 0.042(U_{50} + 0.103U_{30} - 1.415|U_{50} - U_{30}|) + 0.717(R_S + 2.455R_G)$$

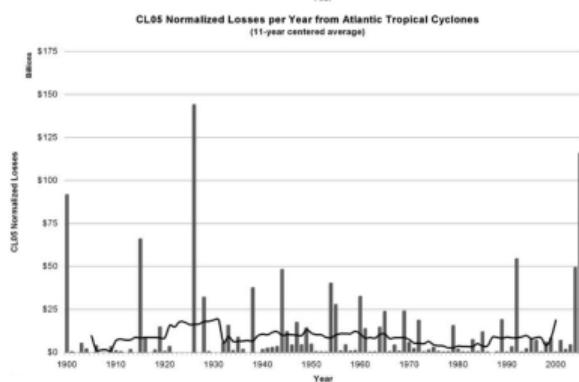
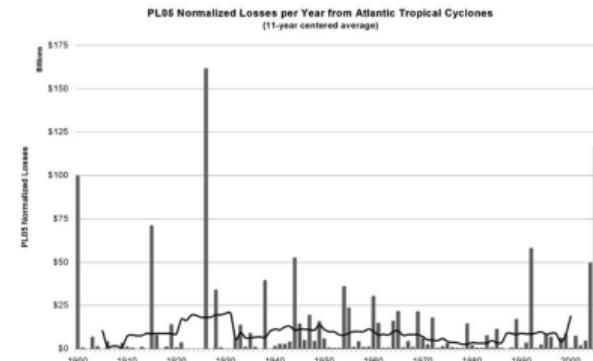
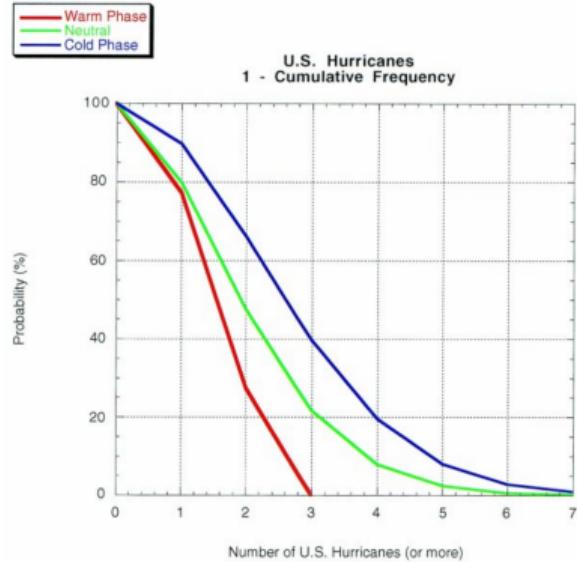
where  $U$ 's are upper-air **zonal winds** at 50 and 30 mb and  $R$ 's are composite functions of August-September western Sahel ( $R_S$ ) and August-November Gulf of Guinea ( $R_G$ ) **rainfall**.

See also [Klotzbach & Bell \(2017\)](#) on the Landfalling Hurricane Probability Project (with all landfall probability calculations)

A	N	O	P	Current-Year Probability	
				(Using Poisson)	(Using Poisson)
3	Climatological Probability				
4	H	MH	H	MH	
5 State					
6 Texas	33%	12%		21%	7%
7 Louisiana	30%	12%		19%	7%
8 Mississippi	11%	4%		6%	3%
9 Alabama	16%	3%		10%	2%
10 Florida	51%	21%		35%	13%
11 Georgia	11%	1%		7%	1%
12 South Carolina	17%	4%		11%	2%
13 North Carolina	28%	8%		18%	5%

## Statistical Perspective

Need (long) historical data,  
(if possible with normalized losses)  
see Pielke Jr. *et al.* (2008)

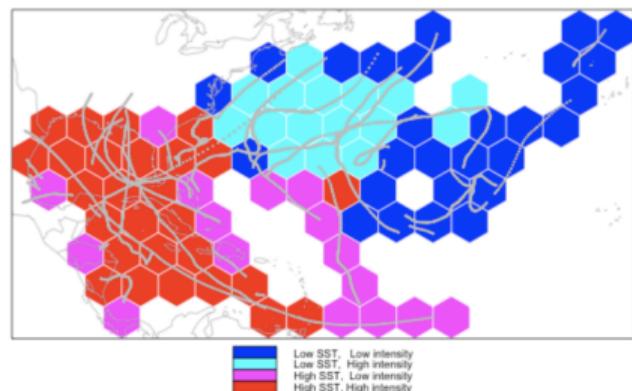
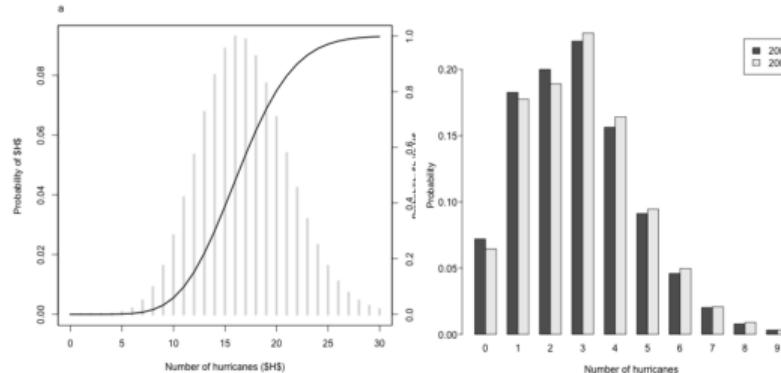
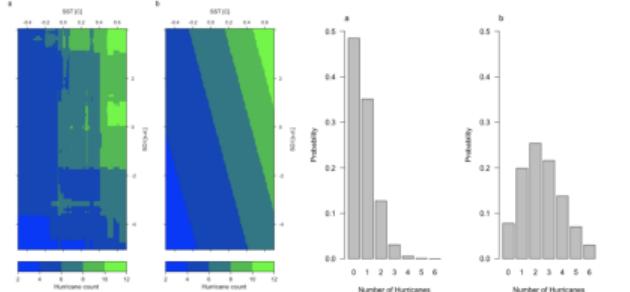


From Bove *et al.* (1998) for hurricanes.

# Loss Probability : Statistical Perspective

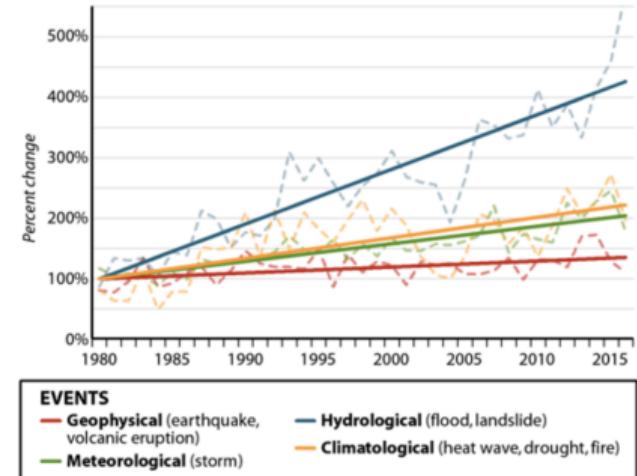
See [Elsner & Jagger \(2013\)](#) for a exhaustive statistical analysis of hurricanes,

- Poisson regression for **counts**
  - Southern Oscillation Index (SOI)
  - sea-surface temperature (SST)
- regression for **intensity**
- spatial models for **trajectories**
- Bayesian models for **one year prediction**



# Loss Probability : Statistical Perspective

GLOBAL TRENDS IN NATURAL CATASTROPHES  
Percentage change each year in number of events compared to 1980



SOURCES: MunichRe NatCatSERVICE; European Academies Science Advisory Council

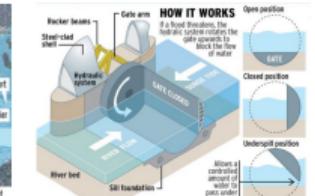
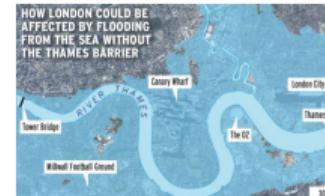
hard to justify in a changing environment

*"In order to apply any theory we have to suppose that the data are homogeneous, i.e. that no systematical change of climate and no important change in the basin have occurred within the observation period and that no such changes will take place in the period for which extrapolations are made"*, Gumbel (1941)

see flood risk (dams, reservoirs, etc)

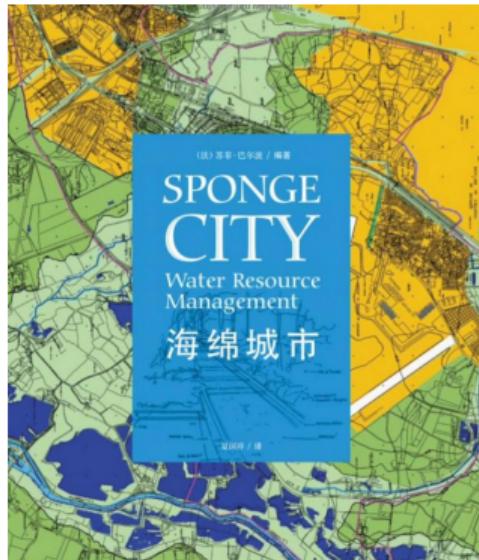
see storms or heatwaves (climate change)

Charpentier (2011)



## The Architect / Urban Planner Perspective ?

See Dong & Han (2011), Bardaux (2016), Li et al. (2017) Jiang et al. (2018) on sponge cities (urban underground water system operates like a sponge to absorb, store, leak and purify rainwater, and release it for reuse when necessary)



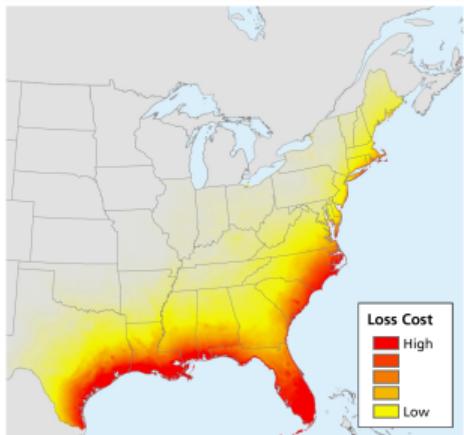
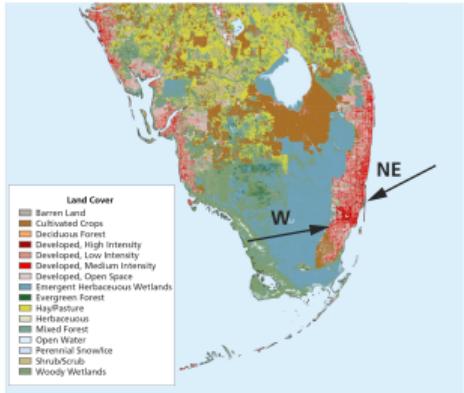
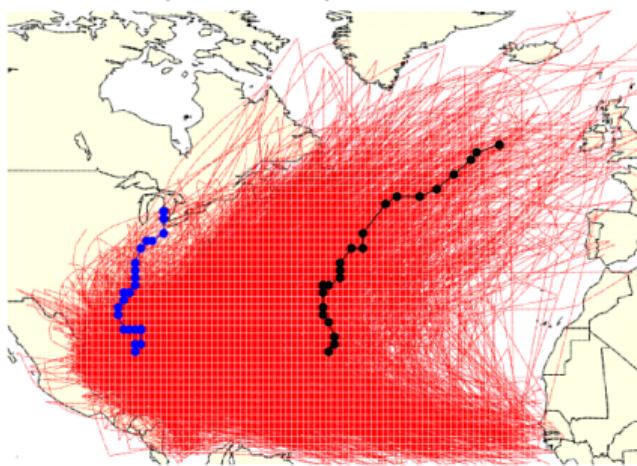
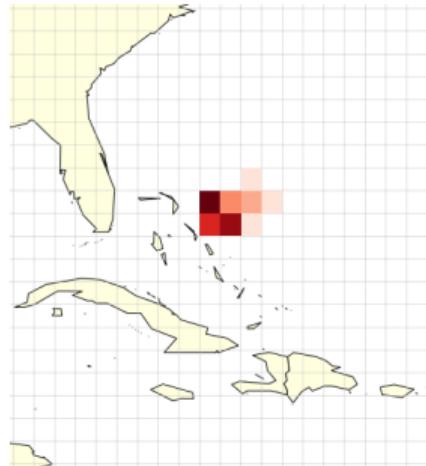
# Loss Probability $p$ : Actuarial Perspective

Actuaries use catastrophe softwares

- Risk Management Solutions ([RMS](#))
- AIR Worldwide ([AIR](#))
- Risk Quantification & Engineering ([RQE](#)) EQECAT

see [Cole, Macpherson & McCullough \(2010\)](#).

Generation of climatic scenarios (+ losses)



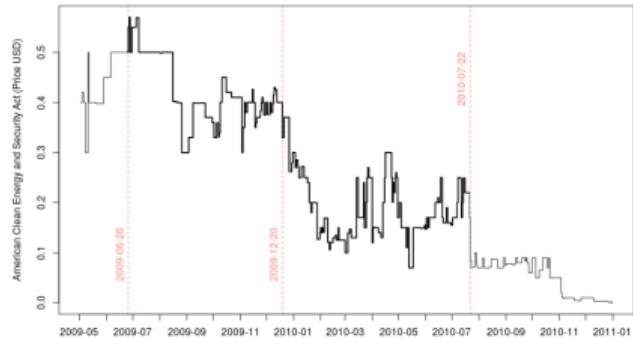
see for Markovian generation [Charpentier \(2014\)](#).

# Predictive Markets

## Peer prediction systems

Wolfers & Zitzewitz (2004) for a description

- elections (e.g. popes, XVIth Century) Rhode & Strumpf (2008)
- politics, Meng (2016) Waxman-Markey bill (2009-2010, finally rejected)
- sports (racetrack bets) Ali (1977)
- infectious diseases, Polgreen, Nelson & Neumann (2006) or Tung, Chou & Lin (2015)
- climate risk Hallstrom & Smith (2005) for hurricanes



## (Perceived) Loss Probability $p$ : Betting Markets

See Eisenberg & Gale (1959) model, on consensus and subjective probabilities (used in Manski (2004) on all-or-nothing contracts)

Individual  $i$  (wealth  $b_i$ ) can bet on horse  $j$  an amount  $\beta_{i,j}$  (so that  $\beta_{i,1} + \dots + \beta_{i,J} = b_i$ ). Assume that  $b_1 + \dots + b_I = 1$ .

Let  $\pi_j$  denote the sum bet on horse  $j$ ,  $\pi_j = \beta_{1,j} + \dots + \beta_{I,j}$ . Observe that from the budget constraint  $\pi_1 + \dots + \pi_J = 1$ , hence  $\pi_j$ 's are probabilities.

Individual beliefs can be related expressed through a probability vector  $\mathbf{p}_i = (p_{i,1}, \dots, p_{i,J})$ , then there is an equilibrium if

$$p_{i,j} = \pi_j \cdot \max_s \left\{ \frac{p_{i,s}}{\pi_s} \right\} \text{ as soon as } \beta_{i,j} > 0$$

(so called Eisenberg-Gale equilibrium).

# Collective Dimension of Insurance (Climate) Risk

*“Insurance is the contribution of the many to the misfortune of the few”*

*“[L'assurance distingue] entre le dommage que subit tel ou tel individu — c'est affaire de chance ou de malchance — et la perte liée au dommage dont l'attribution est, quant à elle, toujours collective et sociale”, Ewald (1986).*

- pricing perspective (zero-sum game, ex-post subsidizing)
- modeling perspective (strongly correlated risks - spatially)
- prevention for climate risks is essentially collective

see Charpentier, Barry & Gallic (2019) for additional thoughts

## Take-Away Conclusion

- knowing **real probabilities** of occurrence disasters is either complicated (hurricanes) or very complicated (flood)
- those probabilities are necessary to assess solvency of insurance companies (central limit theorem is based on true probabilities)
- insurance pricing is based on **beliefs** of insured, and insurance companies
- **predictive markets** can be an interesting revelation mechanism of crowd beliefs (can be used to assess if an insurance market can actually exist, or not)
- it is difficult to think of climate risk without its **collective dimension** ...