

Climate change

- and its impact on building
water damage

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- ▶ Background
- ▶ Data
- ▶ Claims models
- ▶ Predictions



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Background

- ▶ The insurance industry is vulnerable to climate change
 - Non-life insurance: Buildings, cars, ...
 - Life insurance: Altered risks of death and diseases, ...



Bergen, January 2005: Storm surge

Background

- ▶ Back in 1999, Gjensidige Forsikring and Norwegian Computing Center started their fruitful collaboration
- ▶ Three major projects have been conducted to date
- ▶ Focus throughout: Externally inflicted water damage to buildings
- ▶ Overall aim:
 - Admitting that there will be certain climatic changes, what are the level and geographical pattern of losses to be expected in the future?

Background

- ▶ Prediction of future losses calls for:
 - A thorough understanding of the weather mechanisms that influence loss events and their severity
 - Information about future climate
 - Methods to merge the two types of knowledge

Background

- ▶ Relevant questions:
 - Can coherences between losses and weather be quantified?
 - Which weather elements are important?
 - Do coherences differ among different areas?
- ▶ In what ways can an insurance company make use of the results?
 - Limit the effects of climate change through preventive discussions with customers and local regulators
 - Update the company's risk and premium calculations

Data: Insurance data

- ▶ Water losses from Gjensidige Forsikring's own portfolio of Private building policies
- ▶ 10 years of data (1997-2006)
- ▶ Spatial resolution at the level of municipalities
- ▶ Variables:
 - Number of claims and total payment (index-linked) (daily)
 - Population (number of policies) (monthly)

Data: Insurance data

- ▶ Data are *frequency claims*, i.e. rather small losses that occur "often" (as opposed to major catastrophes)
 - Main categories:
 - Water running into basements from above the ground
 - Blocked pipes
- ▶ Not included: Large-claim situations (losses due to flood, storm surge, slide, ...)

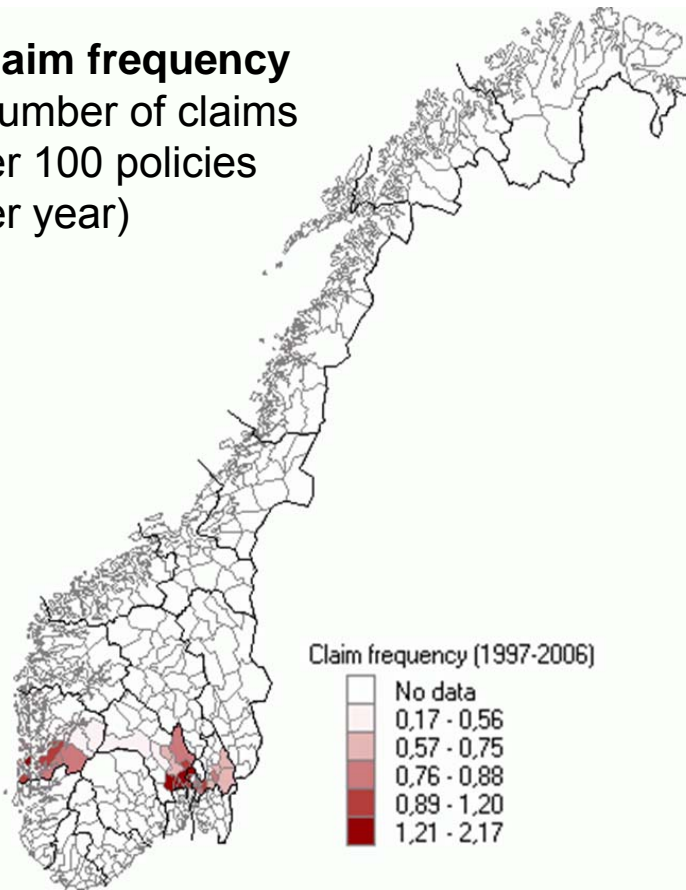


Bergen, September 2005:
Extreme precipitation event

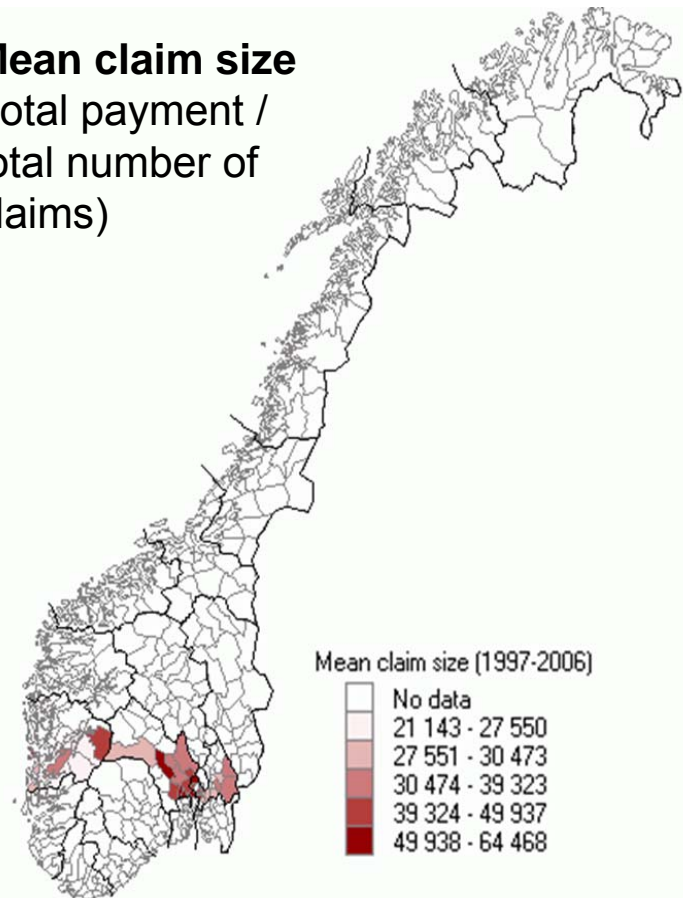
Data: Insurance data

Private buildings, claims from 1997 - 2006

Claim frequency
(number of claims
per 100 policies
per year)



Mean claim size
(total payment /
total number of
claims)

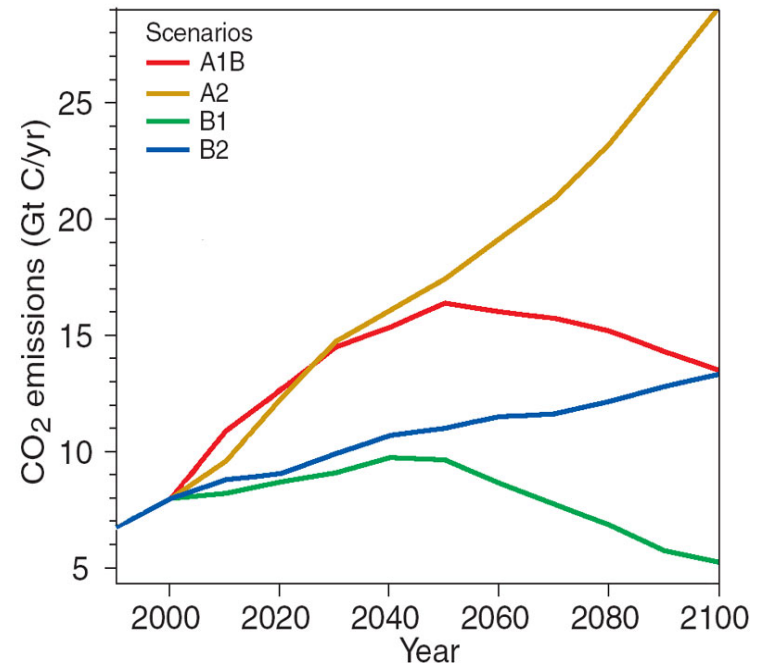


Data: Weather data

- ▶ The Norwegian Meteorological Institute (MI) and Norwegian Water Resources and Energy Directorate (NVE):
 - Daily values for each municipality (1961-2006) based on averages of interpolated (1x1 km²) observations of
 - Precipitation and temperature
 - Runoff and snow water contents
- ▶ Spatial representativity of the values is improved by averaging over the most densely populated areas of a municipality only, *i.e.* the areas where losses will primarily occur
 - Essential since harmful weather often is very local

Data: Climate model data

- ▶ Regionally downscaled and locally adjusted global Hadley Institute HadAM3H model runs under the CO₂ emissions scenarios A2 and B2
- ▶ The same variables available as for the weather data
- ▶ Climate model runs cover two separate periods:
 - 1961 – 1990 (control)
 - 2071 – 2100 (scenario)

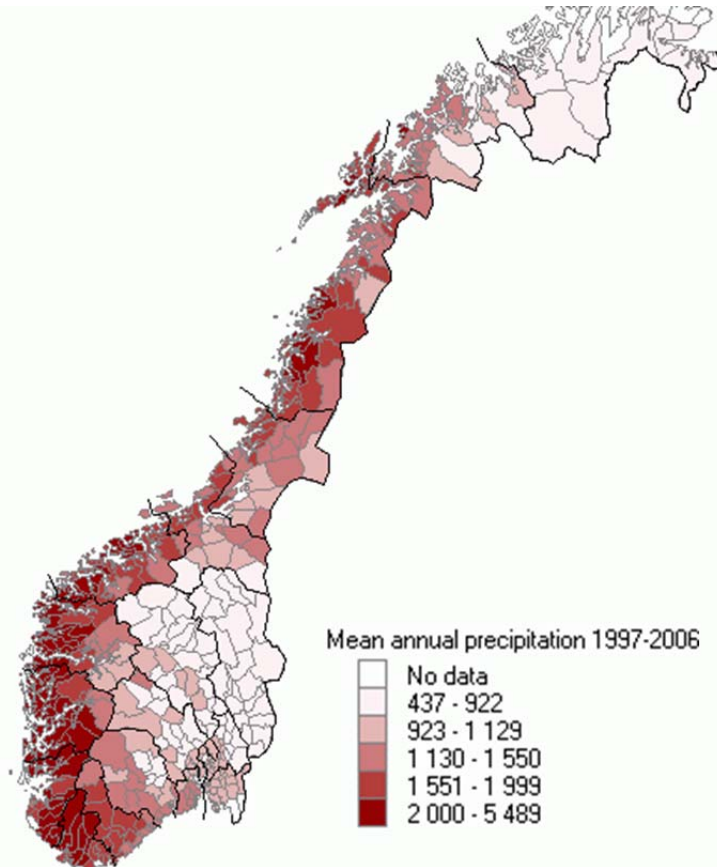


Data: Climate model data

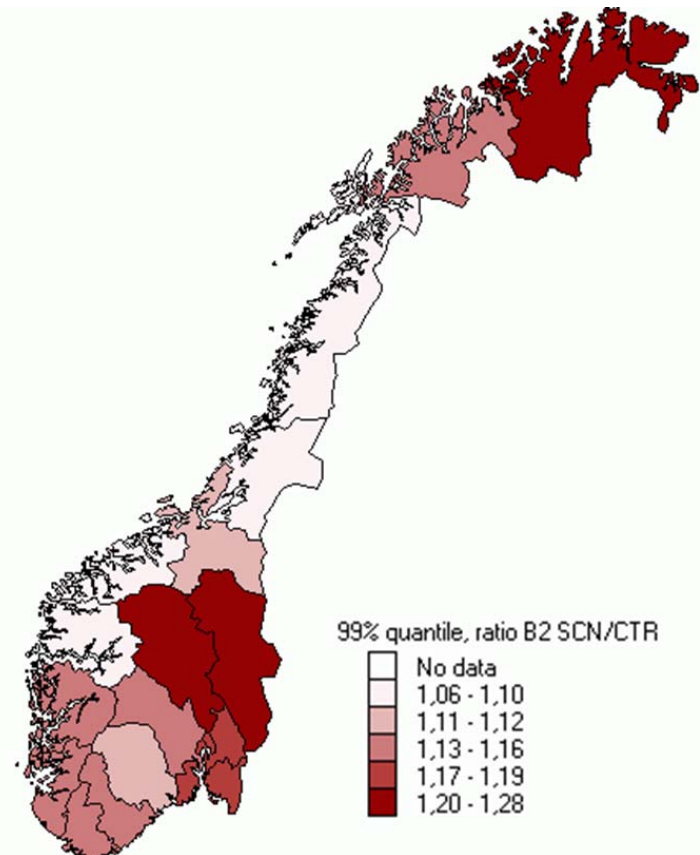
- ▶ Note:
 - Climate model data deviate from observed weather
- ▶ Therefore, climate model data are adjusted locally against weather observations over the control period 1961 – 1990
- ▶ After adjustment, the two data sets agree (locally) at an aggregate level (annual and monthly means and standard deviations)
- ▶ Still, however, variables from the two data sets are not supposed to agree *on a certain day*

Data: Weather and climate data

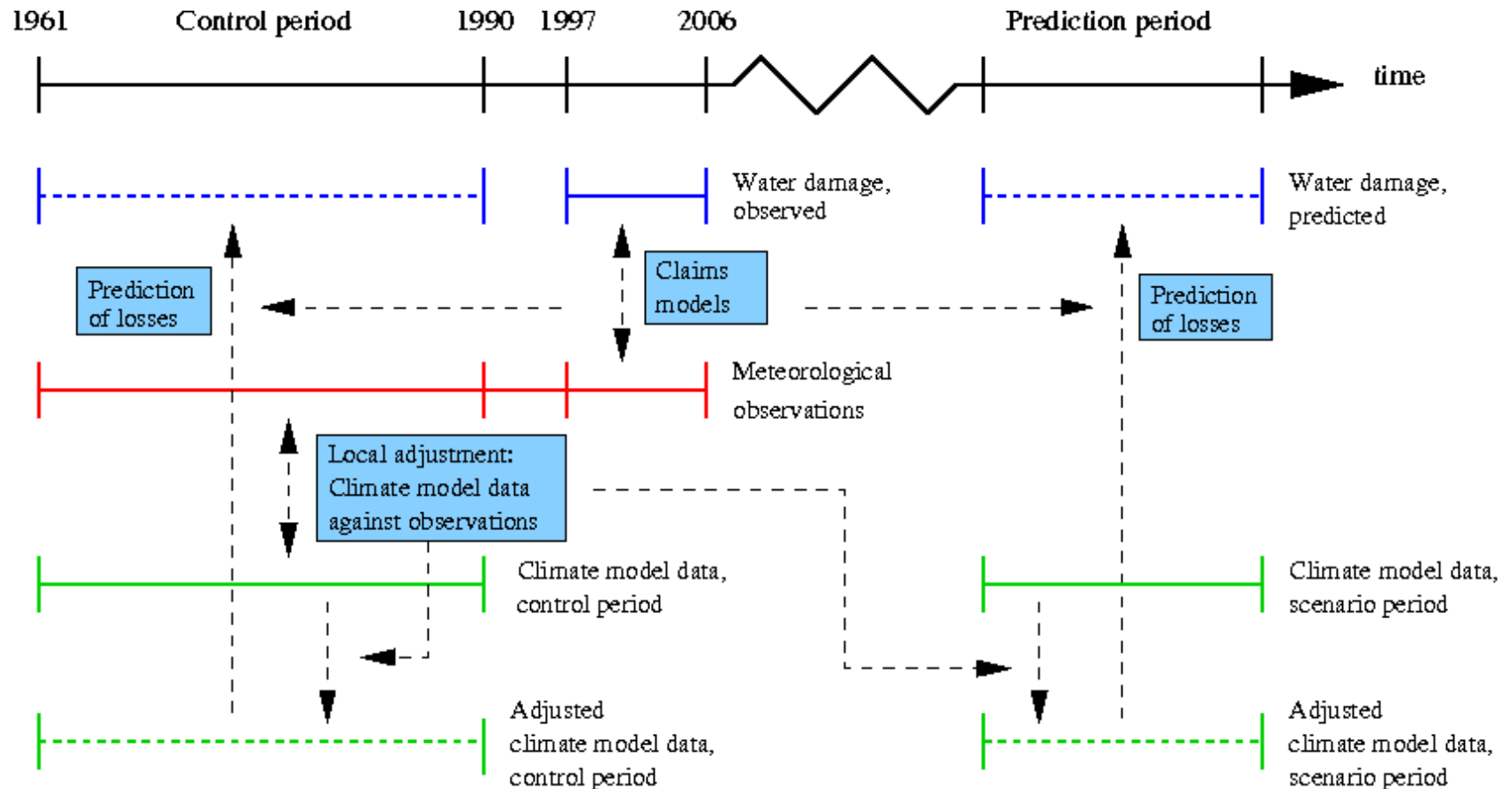
Mean annual precipitation
1997 - 2006



Change in precipitation:
Ratio of 99% quantiles from
B2 scenario and control period



How data are used...



Claims models

What determines the loss level in an area over time?

- ▶ The local climate in the area
- ▶ The sensitivity of the buildings in the area



Claims models

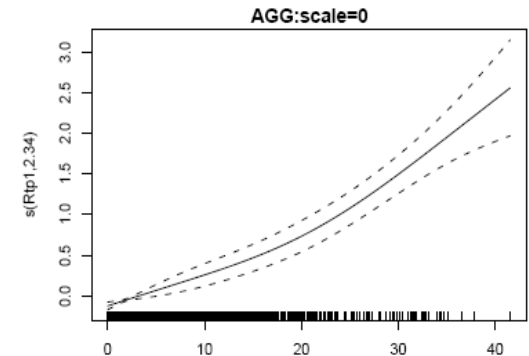
- ▶ Statistical models for the coherence between water damage and daily weather variables established by means of GAM/GLM techniques
- ▶ *Number of claims and mean claim size* modelled separately
- ▶ Number of claims:
 - All days contribute information, also those with no claims
 - Lots of data, large variability:
 - Apparently identical weather conditions may give none or a few claims, *i.e.* not all underlying aspects of the claims are fully described through the weather
- ▶ Mean claim size:
 - Fewer data: Only days with claims contribute information on claim size

Claims models: Explanatory variables

- ▶ Tentatively, all claims day weather variables are included
 - Precipitation, temperature, runoff, snow water contents
- ▶ Additional variables:
 - Precipitation registered the day after the claims day (the precipitation day is from 06UTC – 06UTC...)
 - Aggregated precipitation over the five days before the claims day (saturation of ground and vegetation)
 - Trend: Accounts for potential systematic development of losses over time that is not related to weather (e.g. economic activity)
 - Seasonality: Periodic variation in claims level throughout the year

Claims models

- ▶ GAM plots are used to identify the most likely parametric forms of the explanatory variables to be used in the GLMs
- ▶ GLM model fitting
 - Applied to data from 1997 – 2006
 - Daily claims and weather data are matched at municipality level
 - Identical models throughout (most of) the country (*i.e.* the same explanatory variables are used)



Claims models: Number of claims

- ▶ Modelling the number of claims:
 - Separate model fits for each of 19 counties
 - Individual constant terms for each municipality
 - Other explanatory variables have coefficients that are common to all municipalities within a county



Example

Sogn og Fjordane county:
26 municipalities

Claims models: Number of claims

- ▶ Quasibinomial model, municipality k :

$$N_{kt} \sim \text{quasiBin}(A_{kt}, p_{kt})$$

$$E(N_{kt}) = A_{kt} p_{kt}$$

$$\text{Var}(N_{kt}) = \varphi_k A_{kt} p_{kt} (1 - p_{kt})$$

- ▶ GLM with a logit link applied to the claims probability,

$$\text{logit}(p_{kt}) = \beta_0 + \delta_k + \sum_{i=1}^p \beta_i x_{kti}, \quad \sum_k \delta_k = 0$$

φ_k : Dispersion parameter

N_{kt} : Number of claims, day t

A_{kt} : Number of policies, day t

p_{kt} : Claims probability, day t

β_0 : County component of constant term (common across municipalities)

δ_k : Municipality adjustment of β_0

x_{kti} : The level of explanatory variable i , day t

β_i : Coefficients common across municipalities within a county

Claims models: Claim severity

- ▶ Modelling mean claim size:
 - Separate model fits for each of 5 regions (Gjensidige splits the country into regions with 2 – 5 counties)
 - Individual constant terms for each county
 - Other explanatory variables have common effects in all municipalities within a region

Example

Region "Innland",

2 counties:

- Hedmark
- Oppland



Claims models: Claim severity

- ▶ Gamma model for the mean claim size, municipality k :

$$\bar{S}_{kt} | N_{kt} \sim \text{Gamma}(\xi_{kt}, \nu N_{kt})$$

$$E(\bar{S}_{kt} | N_{kt}) = \xi_{kt}$$

$$\text{Var}(\bar{S}_{kt} | N_{kt}) = \xi_{kt}^2 / (\nu N_{kt})$$

- ▶ GLM with a log link applied to the expected claim size

$$\log(\xi_{kt}) = \alpha_0 + \gamma_{F(k)} + \sum_{i=1}^p \alpha_i x_{kti}$$

$F(k)$: County index

s_{ktj} : Individual claim size, day t

\bar{S}_{kt} : Mean claim size, day t :

$$\bar{S}_{kt} = \sum_{j=1}^{N_{kt}} s_{ktj} / N_{kt}$$

$\bar{S}_{kt} | N_{kt}$: Mean claim size given the number of claims

α_0 : Region component of constant term (common across counties)

$\gamma_{F(k)}$: County adjustment of α_0

x_{ktj} : The level of explanatory variable i , day t

α_i : Coefficients common across counties within a region

Claims models: Total payment

- ▶ Total payment may be derived from the number of claims model and the mean claim size model
- ▶ Expected payment at day t in municipality k is

$$\begin{aligned} E\left(\sum_{j=1}^{N_{kt}} s_{ktj}\right) &= E(N_{kt} \bar{S}_{kt}) \\ &= E(N_{kt}) E(\bar{S}_{kt} | N_{kt}) \\ &= A_{kt} p_{kt} \xi_{kt} \end{aligned}$$

Claims models: Final models

- ▶ Final GLM model selection by means of BIC
- ▶ Tested a subset of reasonable models (potential parametric forms from GAM plots)
- ▶ Final decision based on the sum of BIC-values over all counties or regions

Claims model	Weather-related explanatory variables
Number of claims	Precipitation Runoff Temperature Snow water contents
Claim severity	Precipitation Runoff (partly)

Prediction

- ▶ Claim level predictions for the control and scenario periods are calculated from:
 - Fitted claims models (number and size)
 - Adjusted climate model data for both periods
- ▶ Predictions will be uncertain:
 - Uncertainty due to model fit (quantifiable)
 - Estimated from model simulation
 - Error due to true model discrepancy (not quantifiable)
 - Uncertainty in the climate model data (not quantifiable)

Prediction: Claims change

- ▶ Compare loss level predictions for the control and scenario periods by looking at ratios:

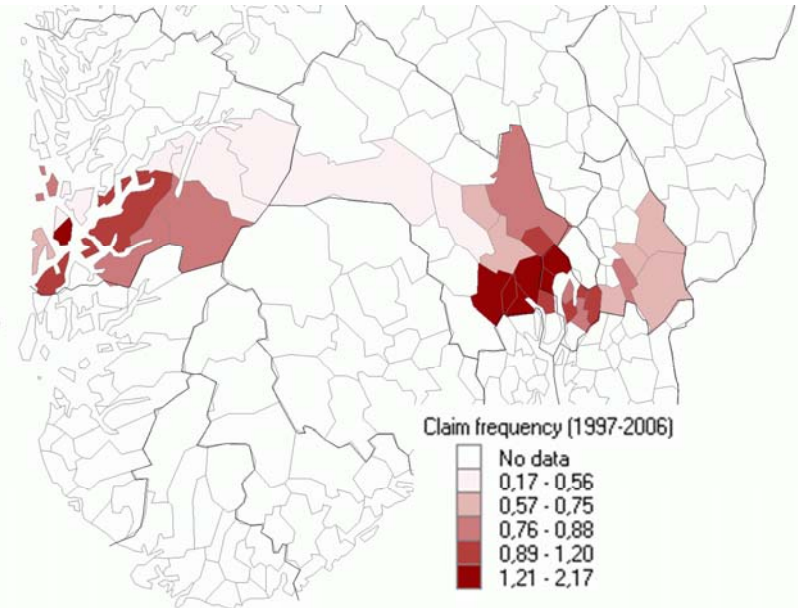
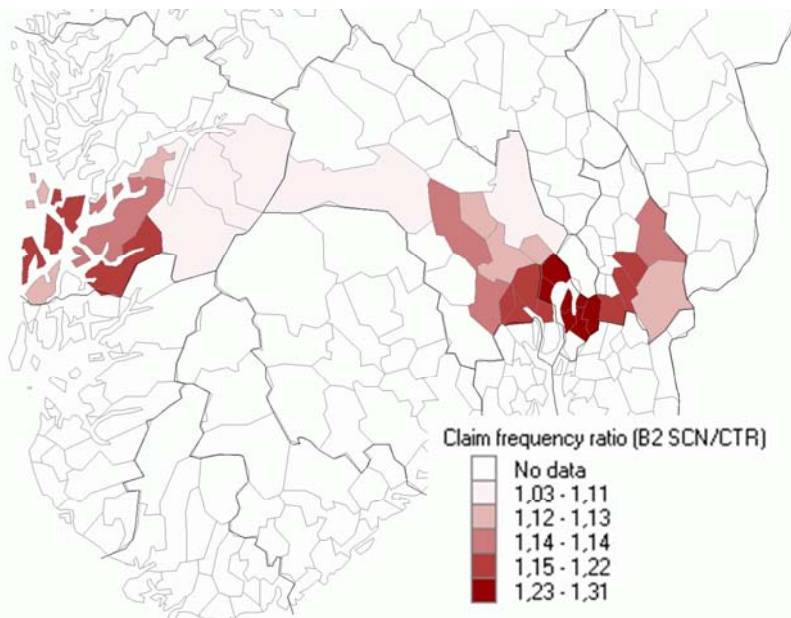
- Number of claims:
$$r_k(scen, ctr) = \frac{\sum_{t \in scen} \hat{N}_{kt}^{scn} | A_{kt} = A_0}{\sum_{t \in ctr} \hat{N}_{kt}^{ctr} | A_{kt} = A_0}$$

- ▶ The ratio for municipality k is a function of
 - The local vulnerability of the buildings (claims model)
 - The climatic change as told by climate models
- ▶ Ratios > 1 indicate an increased future loss level

Prediction: Change in number of claims

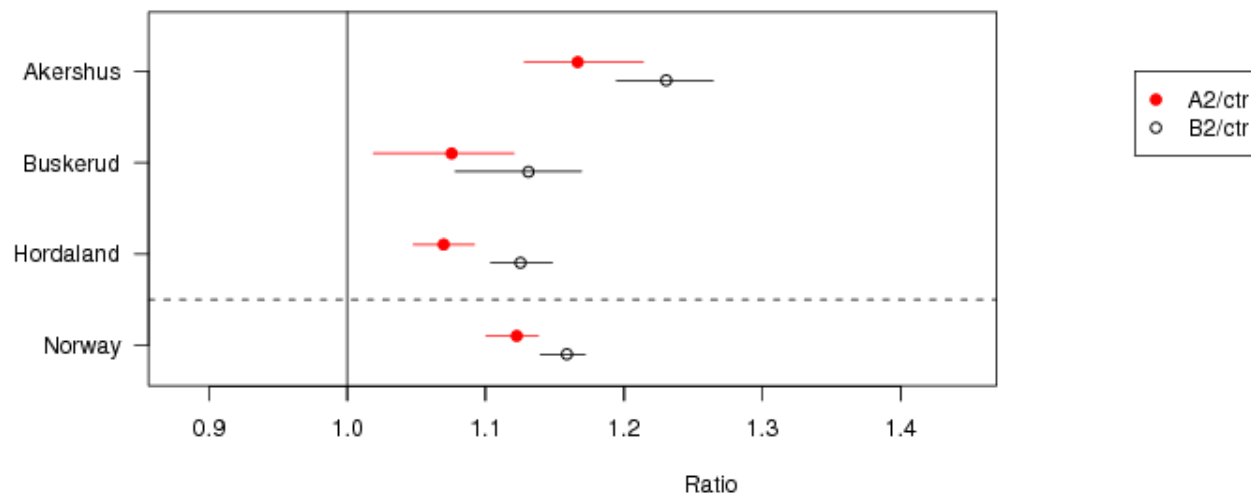
Ratio of number of claims
(scenario period divided
by control period)

Number of claims, 1997 – 2006
(per 100 policies per year)



Prediction: Estimation uncertainty

- ▶ Fitted claims models: Assume a multinormal distribution for the model coefficients
- ▶ Compute a set of ratios from a simulation study ($n=100$)
- ▶ Use empirical quantiles as confidence limits



Summary

- ▶ Future loss level will increase, but predictions show considerable geographical variation
- ▶ Prediction uncertainty is probably substantial, due to:
 - Claims models specification and estimation
 - Climate model data
- ▶ Still, rough estimates indicate that the insurance industry may benefit from taking pre-cautionary measures to reduce undesirable effects of climate change
- ▶ Further research will be carried out
 - EU (Marie-Curie) granted project (2008-2011): *Climate Change and Insurance Industry (CCII)*, together with London School of Economics, Lloyds and Gjensidige Forsikring

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