1 # Comparing an insurer's perspective on building damages with
2 # modelled damages from pan-European winter windstorm event
3 # sets: a case study from Zurich, Switzerland

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9 **Abstract**

10 With access to claims, insurers have a long tradition of being knowledge leaders on damages caused by e.g.

11 windstorms. However, new opportunities have arisen to better assess the risks of winter windstorms in Europe

12 through the availability of historic footprints provided by the Windstorm Information Service (Copernicus

13 WISC). In this study, we compare how modelling of building damages complements claims-based risk

14 assessment. We describe and use two windstorm risk models: an insurer's proprietary model and the open

15 source CLIMADA platform. Both use the historic WISC dataset and a purposefully-built, probabilistic hazard

16 event set of winter windstorms across Europe to model building damages in the canton of Zurich, Switzerland.

17 These approaches project a considerably lower estimate for the annual average damage (CHF 1.4 million),

18 compared to claims (CHF 2.3 million), which originates mainly from a different assessment of the return period

19 of the most damaging historic event Lothar/Martin. Additionally, the probabilistic modelling approach allows

20 assessing rare events, such as a 250-year return period windstorm causing CHF 75 million damages, including

21 an evaluation of the uncertainties. Our study emphasises the importance of complementing a claims-based

22 perspective with a probabilistic risk modelling approach to better understand windstorm risks. The presented

23 open source model provides a straightforward entry point for small insurance companies.

24 ## 1   Introduction

25 Severe windstorms are responsible for widespread socio-economic impacts such as damage to buildings,

26 structures, transport networks, forests, and even loss of lives. Windstorms represent one of the most damaging

27 natural hazards in many parts of the world, not least in Switzerland (Imhof, 2011). In the densely populated

28 canton of Zurich, which is located in north-eastern Switzerland, windstorms are among the most destructive

29 natural hazards: building damages due to windstorms amount to 30 % of the total amount of building damage

30 from natural hazard in this region. For comparison, damages due to hailstorms and flooding amount to 41 % and

31 28 %, respectively (all numbers from 2018; GVZ annual report, 2018; Schadenstatistik VKG, 2020).

32 In general, the impact of a windstorm in terms of building damages depends on the severity of associated

33 surface winds and gusts as well as on the exposed values and the respective vulnerability (i.e., damage

34 susceptibility) of the buildings being subject to the hazard − with both building stock and vulnerability changing

35 over time. High wind speeds cause large pressure and suction effects, which in turn are responsible for damage

36 to the roof and the building facade. Damaging winds and violent gusts in the canton of Zurich are mainly due to

37 the passage of large-scale extratropical cyclones and their associated fronts during autumn and winter as well as

38 due to mostly local convective storms during summer. Winter windstorms typically cause widespread minor

39 building damages summing up to large amounts, whereas it is not unusual that summer convective storms cause

40 major damages of only a few buildings due to locally very high wind speeds.

41 The cantonal building insurance GVZ compulsorily insures all buildings in the canton of Zurich (with a few

42 exceptions) against damage due to natural hazards and fire: i.e. in total around 300'000 buildings with a total

43 sum insured of around Swiss Francs (CHF) 500 billion (in 2018). GVZ is an independent institution of the

44 canton of Zurich under public law (GVZ homepage, 2020).

45 Windstorm damage events in the canton of Zurich have been recorded in GVZ's database since 1981. For

46 example, windstorm Lothar on 26 December 1999 caused total insured building damages of around

47 CHF 60 million and is by far the most extreme windstorm event in the database. Second largest is windstorm

48 Burglind on 3 January 2018 (Scherrer et al., 2018), which caused total insured building damages of more than

49 CHF 14 million. The most extreme summer damage event in GVZ's record is due to a very local, but extremely

50 intense convective storm on 2 August 2017 with measured maximum gusts of more than 180 km/h in the

51 lowlands, which caused total insured building damages of approximately CHF 4 million. Even though small-

52 scale convective storm events are potentially hazardous, in this study we focus on large-scale winter windstorms

53 only, which were responsible for around three quarters of all insured windstorm damages in the canton of Zurich

54 since 1981.

55 Extreme damage events such as those caused by Lothar or even stronger windstorms are rare by definition. For

56 risk assessment, solid estimates of the probability of occurrence of such events are absolutely essential and

57 GVZ's claims data of almost 40 years provides a too short observational period which leads to a large sampling

58 uncertainty. A larger sample of events is needed for which at least quantitative meteorological data and if

59 possible damage data at ideally high spatiotemporal resolution are available (e.g., Haas and Pinto, 2012).

60 Observational damage data are generally sparse and incomplete for historic windstorms in Switzerland (Stucki

61 et al., 2014). Instead, societal actors often use modelled impacts to manage their risk. Insurance and reinsurance

62 companies apply impact models for their pricing and governments use modelled risk for option appraisal (e.g.,

63 The Economics of Climate Adaptation Working Group, 2009; Bresch, 2016). Additionally, the information is

64 needed for climate-related financial disclosure (Surminski et al., 2020). However, only very few impact models

65 are available open source and free access for users both in the scientific as well as public or private domain.

66 Typically, risk is modelled as a combination of hazard, vulnerability, and exposure (IPCC, 2014). The hazard

67 part is the best understood and research culminated in open datasets of historic windstorm events (Roberts et al.,

68 2014; WISC products, 2019), whereas maximum wind gust speeds are frequently used as hazard component to

69 assess windstorm risk (e.g., Klawa and Ulbrich, 2003). Vulnerability has been covered by many studies and

70 reviews (e.g., Della-Marta et al., 2010; Schwierz et al., 2010; Feuerstein et al., 2011; Prahl et al., 2015; Koks

71 and Haer, 2018). There are many theoretical learnings from these studies, but an implementation in a

72 comprehensive open source and easy access risk assessment model is still missing. Detailed exposure data are

73 generally not publicly available and many societal actors have their own detailed view on exposure and do not

74 need to rely on a publicly available dataset. There are open, spatially explicit datasets available based on the

75   distribution of nightlight and population (Eberenz et al. 2019), based on the Gross Domestic Product (GDP;

76   Geiger et al. 2018), or on building data from OpenStreetMap (Koks and Haer, 2018). The sparse availability is

77   why in some research studies loss ratios were used instead of information on exposure (Donat et al., 2011).

78   Using the modelling approach for Switzerland, Welker et al. (2016) applied the methods presented first by

79   Stucki et al. (2015) to a sample of more than 80 high-impact winter windstorms that affected Switzerland in

80   1871-2011. The approach involves the dynamical downscaling of the Twentieth Century Reanalysis (20CR)

81   using the Weather Research and Forecasting (WRF) model. The calculated windstorm footprints served as input

82   for the modelling of economic damages using a precursor of the open source impact model CLIMADA

83   (CLIMate ADAptation; Aznar-Siguan and Bresch, 2019a). CLIMADA was successfully applied in several other

84   studies for the purpose of risk assessment and quantification of socio-economic impacts (e.g., Della-

85   Marta et al., 2010; Schwierz et al., 2010; Raible et al., 2012; Reguero et al., 2014; Gettelman et al, 2018; Walz

86   and Leckebusch, 2019).

87   To increase the sample of windstorm footprints available for risk assessment, insurance and reinsurance

88   companies often combine observed windstorm footprints as far as available with synthetic footprints generated

89   by stochastic or dynamic atmospheric models. In this way, they obtain a more comprehensive view on risk.

90   The Windstorm Information Service (WISC) of the Copernicus Climate Change Service aims to provide a

91   consistent and open database of hazard data to assess the risk of windstorms in Europe for all kinds of players in

92   the insurance sector and beyond. The centrepiece of the WISC dataset are wind gust footprints at high spatial

93   resolution of approximately 4.4 km for, on the one hand, a historic hazard event set of around 140 European

94   winter windstorms in 1940-2014 and, on the other hand, a synthetic hazard event set of around 23'000 events.

95   Similar to the predecessor project Extreme Windstorms Catalogue (XWS; Roberts et al., 2014), the WISC

96   historic hazard event set contains windstorms that hit Europe, but provides the corresponding wind gust

97   footprints at improved spatial resolution and covers more windstorms over a period longer than the claims

98   database available to most insurance companies. This makes it possible to reduce the sampling uncertainty of

99   the risk assessment. The windstorm hazard event sets as provided by WISC form an independent database to

100   validate and further develop existing European winter windstorm models. The dataset can be used for both pan-

101   European analyses and local analyses, as shown in this study.

102   Using the WISC historic hazard event set allows GVZ in a way to "re-check" historic events. By means of the

103   synthetic hazard event set, the tail of the hazard and damage distributions should be investigated. However,

104   Röösli et al. (2018) found that the synthetic hazard event set is not suitable for this purpose. Therefore, we

105   propose instead a probabilistic windstorm hazard event set based on a method described in

106   Schwierz et al. (2010) to overcome the shortcomings of the WISC synthetic hazard event set. This new

107   probabilistic hazard event set of around 4'300 events contains windstorms from the WISC historic hazard event

108   set altered by various perturbations. As discussed in this study, such a statistical perturbation is based on the

109   same observational period as the WISC historic hazard event set and therefore cannot reduce the sampling

110   uncertainty.

111   This study shows how GVZ uses both the WISC dataset and the new probabilistic hazard event set for assessing

112   the potential building damage and risk due to extreme windstorm events, including an evaluation of the

113   uncertainties of such assessments. A relationship between wind gust speed in the affected region of the canton

114   of Zurich and associated building damages is found, which allows for a rapid, straightforward estimation of

115   damage directly after the occurrence of extreme, unprecedented windstorms. This study further shows how GVZ

116   was able to improve its windstorm risk assessment on the basis of the WISC dataset and the new probabilistic

117   hazard event set, and could serve as an example for other players in the insurance sector or other societal actors

118   in Switzerland and in the rest of Europe. At the same time, this study also illustrates selected limitations of the

119   WISC dataset.

120   2   Data and methods

121   After a description of the insurance claims data (Sect. 2.1) and the windstorm hazard event sets used (Sect. 2.2),

122   we introduce the GVZ and the CLIMADA risk assessment models applied for damage modelling (Sect. 2.3) and

123   conclude this section with a brief recapitulation of the risk assessment metrics employed in this study

124   (Sect. 2.4).

125   2.1   Insurance claims data

126   The windstorm damages of past events are recorded in a proprietary database of GVZ. It consists of almost

127   40 years of insurance claims data, in total more than 84'000 single wind damage records. From this database all

128   the events relevant for this study were selected by following the event definition of the windstorm event set

129   "WISC historic" (Sect. 2.2.1). In total, 18 events are associated with WISC windstorms based on that definition

130   (see also Table 1). Due to the nature of the database, only the damage reports actually insured by GVZ were

131   considered. The insurance claims data allow GVZ to assess the risk for its own portfolio by analysing frequency

132   and severity of past damages, i.e. to assess its risk due to winter windstorm events with a return period smaller

133   than 40 years. Additional information can help GVZ to put their recorded damages into reference and to get a

134   better estimate of the risk of events with a return period larger than the 40 years of experience.

135   For the sake of comparability, the insured damages had to be normalised to present-day exposure levels. In this

136   study, the applied normalisation considers the general inflation on the basis of the Zurich construction price

137   index (2020). Hereinafter, both insured and modelled windstorm damages are including occasional deductibles

138   − so-called "gross damages", to ease comparison.

139   2.2   Windstorm hazard event sets

140   Atmospheric models provide information about winter windstorm events that can be used as hazard component

141   in a risk assessment model. WISC published several hazard datasets each containing a set of windstorm events

142   and providing the maximum wind gust per geographic location per event. We used the historic windstorm

143   footprints (Sect. 2.2.1) and constructed a probabilistic extension based on it (Sect. 2.2.3). In addition, we derived

144   wind gust footprints from measurements for a selection of present windstorm events (Sect. 2.2.4). The additional

145   windstorm hazard event sets published by WISC, that are however not considered in this study, are briefly

146   summarised in Sect. 2.2.2.

2.2.1   Historic windstorm hazard event set

The historic windstorm hazard event set − denoted "WISC historic" − contains wind gust footprints for around 140 winter windstorm events in Europe in 1940-2014 (i.e., 75 modelled years in total). The events were selected, on the one hand, based on the high damage they caused and, on the other hand, because of their high intensity in meteorological terms (i.e., high vorticity). Because of this pan-European perspective, the dataset is not necessarily specific to windstorms in the canton of Zurich. Nevertheless, the high-impact windstorms Lothar/Martin (26–28 December 1999) and other intense windstorms such as Vivian/Wiebke (26 February– 1 March 1990) are included.

The windstorm footprints were computed by running the UK Met Office Unified Model (MetUM; Davies et al., 2005) at approximately 4.4 km resolution with ERA-20C reanalysis (Poli et al., 2016) and ERA-Interim reanalysis (Dee et al., 2011) as boundary conditions, covering Europe and parts of the North Atlantic. ERA-20C was used for all windstorm events in 1940-1979 and ERA-Interim for all events in 1979-2014.

Each of the footprints is composed of gridded maximum 3-second gusts, with maxima determined for a 72-hour time window. This relatively long time window was chosen, because it is widely used in the insurance sector (WISC products, 2019). However, it also implies that the footprints of directly successive events (i.e., with a time difference of less than 72 hours) such as Lothar (26 December 1999) and Martin (27-28 December 1999) are combinations of the footprints of both successive events. In this study, the WISC windstorm footprints for events that have overlapping time windows are combined to represent one event – as insurance claims data does often not represent the exact time/date of damage either (for various reasons, a key one being reporting uncertainties). This combination is necessary to make sure that a maximum that occurred only once (e.g., the wind gusts reached during Lothar) is only represented once in the hazard event set (as event Lothar/Martin) and is not represented twice (once as Lothar and once as Martin). There are five pairs of windstorms with overlapping time windows in the original dataset that were combined by taking the maximum wind gust of both footprints at each location, giving in total 142 windstorm events (Table 1). The problem of overlapping windstorm footprints and the resulting combination of events could have been prevented by incorporating the geographical information into the event definition. For example, Roberts et al. (2014) aggregated only the wind gusts within a certain radius around the windstorm centre into a footprint to avoid this problem.

The wind gust speeds from "WISC historic" are considered to be realistic compared to observations for areas at sea level (WISC products, 2019). However, with regard to the hilly topography of the canton of Zurich the question arises as to how realistic the underlying model topography is in comparison to the real topography and, as a result, how good the height-dependent wind gust speeds are compared to observational data. Even though this could not be finally clarified in this study since available wind measurements are generally too sparse for historic windstorms in the canton of Zurich, a correction of all the WISC wind gusts in the form of simple correction factors does not seem reasonable and was therefore not applied.

2.2.2   Other WISC hazard event sets

There are two additional windstorm hazard event sets published by WISC, that are however not analysed in detail in this study:

184 1. The operational windstorm hazard event set − denoted "WISC operational" − contains around

185 110 windstorm events in 1979-2017 and thus more recent events than the windstorm hazard event set "WISC

186 historic" used in this study, which contains windstorm events until 2014 only. "WISC operational" is based on a

187 new generation of atmospheric reanalysis, the ERA5 reanalysis (Hersbach and Dee, 2016). As it does not cover

188 the time range 1940-1979 (compared to "WISC historic") it does not complement the recorded damages by

189 providing information about historic events not covered by GVZ's claims database.

190 2. The synthetic windstorm hazard event set − denoted "WISC synthetic" − was created within the

191 UPSCALE (UK on PRACE - weather-resolving Simulations of Climate for globAL Environmental risk;

192 UPSCALE, 2020) modelling framework and is a physically realistic set of plausible winter windstorm events in

193 the period 1985-2011 based on the climatic conditions of that period. The modelling framework developed five

194 ensembles. The dataset contains wind gust footprints for around 23'000 synthetic windstorms: i.e., three sets of

195 7'660 events each. Each of the three sets covers 135 modelled years. The original idea of the hazard event set

196 "WISC synthetic" was to use wind information from climate models to provide wind gust footprints for winter

197 windstorms in Europe with a return period of 250 years or even higher. However, this hazard event set was not

198 considered because the findings of Röösli et al. (2018) could be replicated in this study, showing that the dataset

199 does not contain the maximum wind gust speeds we would expect from the distribution of the historic

200 windstorm hazard events (Fig. A1) nor the high intensities we would expect from very rare, high-impact

201 windstorm events (Fig. 1).

202 For a detailed description of all unused windstorm hazard events sets provided by WISC, we refer to the

203 documentations available online at WISC products (2019) and WISC hazard event set description (2019).

2.2.3 Probabilistic windstorm hazard extension

205 Based on "WISC historic", we generated an additional probabilistic windstorm hazard event set − denoted

206 "WISC probabilistic extension". By applying a method described in Schwierz et al. (2010), the individual

207 windstorm events in "WISC historic" (parent events) were altered to create 29 altered offspring events by

208 various perturbations: e.g., spatial displacement and by weakening / intensifying the wind speeds (non-altered

209 wind speeds are spatially displaced only). The spatial displacement was undertaken by shifting the respective

210 windstorm footprint by about 20 km to the north, south, west, or east. The wind gust speeds were intensified and

211 weakened by no more than 3 m/s (normally much less) according to the probabilistic alteration of wind speeds

212 in Eq. (1), with a scale parameter $\alpha = 0.0225$ and a power parameter $\beta = 1.15$ (choice explained further below):

$$windspeed_{scenario\ 1} = windspeed_{original} + \alpha * windspeed_{original}^{\beta}$$

$$windspeed_{scenario\ 2} = windspeed_{original} - \alpha * windspeed_{original}^{\beta}$$

$$windspeed_{scenario\ 3} = windspeed_{original} + \alpha * \sqrt[\beta]{windspeed_{original}} \tag{1}$$

$$windspeed_{scenario\ 4} = windspeed_{original} - \alpha * \sqrt[\beta]{windspeed_{original}}$$

$$windspeed_{scenario\ 5}$$
$$= windspeed_{original} - \frac{\alpha}{2} * windspeed_{original}^{\beta} - \frac{\alpha}{2} * \sqrt[\beta]{windspeed_{original}}$$

213   These newly created "probabilistic" footprints can be viewed as scenarios of plausible windstorms as they only

214   differ slightly from historic events, retaining both the spatial extent and general structure. In countries close to

215   the sea or with a pronounced and high topography, the methodology for creating the probabilistic events might

216   need adaptation to better incorporate the difference in surface roughness and altitude.

217   For using the scenarios in a qualitative risk assessment framework, the probabilistic windstorm footprints can be

218   used as they are, but for a quantitative risk assessment the frequencies of the windstorm footprints need to be

219   estimated. In an effort to assign reasonable frequency estimates to the probabilistic windstorm footprints, we

220   considered the distribution of the historic, pan-European Storm Severity Index (SSI; formula used by Dawkins

221   et al., 2016; further information in Lamb and Frydendahl, 1991; Leckebusch et al., 2008). Similar as in Schwierz

222   et al. (2010), the algorithm of creating the probabilistic windstorm footprints was configured to recreate the

223   cumulative distribution function of a generalized extreme value (GEV) distribution fitted to the historic SSI

224   values. We defined the frequency of all probabilistic windstorm footprints to be equal and to sum up to the

225   frequency of the parent windstorm. We then selected a set of parameters for weakening and intensifying the

226   wind speeds (parameters α and β in Eq. (1)) that resulted in a similar probabilistic distribution of SSI as the

227   extrapolated distribution from the historic SSI values. For the probabilistic hazard event set to best represent the

228   tail of the historic distribution, we determined a combination of α and β, that minimises the difference in the

229   cumulative distribution functions for events that have a return period of > 75 years.

230   "WISC probabilistic extension" includes footprints for 4'118 probabilistic windstorm events, along with the

231   142 original windstorm events in "WISC historic" (Table 1), and provides a basis of an event-based risk

232   assessment for winter windstorms with return periods of around 250 years, a scenario relevant for regulatory

233   requirements in the insurance sector. It is important to note that this method incorporates a lot of uncertainty,

234   including but not limited to the sampling uncertainty of rare events in a relatively short time range (i.e., 75 years

235   in case of "WISC historic").

236   Encouragingly, the hazard event set "WISC probabilistic extension" shows considerably higher wind gust

237   speeds in the canton of Zurich as compared with "WISC synthetic" (Fig. 1). Nonetheless, the maximum wind

238   gust speeds of the most extreme event in "WISC probabilistic extension" are not considerably higher than those

239   of Lothar/Martin, the most extreme event in both "WISC historic" and the insurance claims data.

2.2.4   Observed footprints for current windstorms

241   Real-time wind gust observations can serve as the hazard part of the damage model for a rapid damage

242   estimation directly after the occurrence of an extreme windstorm event. Such "observed" windstorm footprints

243   can also be used for further validation of GVZ's damage modelling approach (Sect. 2.3). To create such

244   footprints, we used interpolated wind gust measurements in the canton of Zurich based on the Common

245   Information Platform for Natural Hazards (GIN; GIN platform, 2019) for a selection of seven winter

246   windstorms in the years 2017 and 2018. With the exception of winter windstorm Burglind hitting Switzerland

247   on 3 January 2018, the windstorms considered caused only minor damages in the canton of Zurich. The

248 individual windstorm footprints are based on a total of around 110 measurement stations in the canton of Zurich

249 and in the immediate vicinity (i.e., buffer zone with a distance of 20 km around the polyline of the canton). For

250 spatial interpolation, we applied an Inverse Distance Weighting (IDW) interpolation with the Shepard method

251 used for weight calculation. In this study, the gridded wind gust footprints derived from measurements have a

252 horizontal resolution of 2 km. The topography of the canton of Zurich is not considered in the applied

253 interpolation method and unquestionably the quality of the derived windstorm footprints could be improved by

254 using a more elaborate interpolation method, which takes account of the topography.

255 **2.3   Damage modelling approaches**

256 The windstorm footprints of the different hazard event sets described in the previous section were used as input

257 for damage modelling and GVZ's proprietary windstorm damage model was applied for this (Sect. 2.3.1). In

258 addition, the CLIMADA impact model was used to be able to publish the method used in this study with open

259 data and open source code (Sect. 2.3.2).

260 In both damage models, the extent of damage results from the intensity of the windstorm event (i.e., hazard), the

261 value of the asset (i.e., exposure), and the susceptibility of the asset to damage (i.e., vulnerability). This concept

262 is broadly used and is explained in more detail in Aznar-Siguan and Bresch (2019a). In this study, the

263 windstorm hazard assessment is based on the winter windstorm footprints described in Sect. 2.2. The exposure

264 is the value of the buildings in the canton of Zurich and the vulnerability is described by a functional

265 relationship that defines how much the buildings are damaged at a certain wind gust speed. In both damage

266 models, we use the vulnerability curve of Schwierz et al. (2010). This vulnerability curve combines the damage

267 degree and the percentage of assets affected. Only damage to buildings is estimated. The estimate does not

268 include damage to movable property, damage to infrastructure, nor business interruption.

269 **2.3.1   GVZ damage model**

270 The damage estimates in this model are computed using a rather conventional modelling framework and the

271 reduced complexity of the approach allows a well interpretable assessment of the model skill. Normally, GVZ

272 uses its damage model directly after the occurrence of a windstorm event to estimate the expected building

273 damage. Furthermore, GVZ applies the damage model to estimate the damage potential and the risk associated

274 with windstorms with regard to solvency considerations and prevention options. The main points of the

275 modelling approach are described in the following.

276 The initial step is a simple spatial overlay of the gridded maximum wind gust speeds during the respective

277 windstorm event with GVZ's current building stock (from 2018; without sublevel garages, as they are usually

278 not affected by windstorms), where GVZ's proprietary building database with information about e.g. the sum

279 insured of each building and the publicly available building footprints (GIS browser Zurich, 2019) were used.

280 GVZ's insurance penetration in the canton of Zurich is almost 100 %. In the damage model, damage is possible

281 from a wind gust speed of more than 90 km/h, and only buildings affected by such gusts were considered in the

282 following modelling steps.

283 Figure A2 shows the spatial distribution of all insured buildings in the canton of Zurich as well as of the total

284 sum insured at municipal level. The aggregated sum insured for all buildings in the two main cities, Zurich and

285 Winterthur (municipal boundaries indicated by blue polygons), accounts for almost 40 % of the total insured
286 value for the entire canton.

287 To estimate the damage in monetary terms, the value of each individual building (i.e., its insured value) was
288 multiplied by the factor "Mean Damage Degree" (MDD, a number between 0 and 1) calculated from the
289 vulnerability curve of Schwierz et al. (2010), where the gust speeds at building level computed in the first step
290 were converted into the corresponding MDD factors. The MDD factors are a non-linear function of the
291 maximum wind gust speed during a windstorm event and are diagrammed in Welker et al. (2016). The same
292 vulnerability curve of Schwierz et al. (2010) is also implemented in the open source impact model CLIMADA
293 (Aznar-Siguan and Bresch, 2019a).

294 In the next step of the damage model, the probability of buildings affected is calculated with a stochastic
295 approach. The respective windstorm event was automatically categorised according to its severity (here,
296 according to the 95th percentile of all gust speeds at building level in the affected region of the canton of
297 Zurich), from which the assumed degree of impact is derived. The degree of impact for the different windstorm
298 categories (i.e., a percentage of total affected buildings for the canton of Zurich, m) was derived from
299 proprietary event damage data from GVZ's database. Then, a random sample of m buildings was selected, with
300 the number m depending on the windstorm's severity. Only buildings with MDD > 0 were considered, i.e. only
301 those buildings with potential damage > 0. For the selected buildings, the amount of damage at building level
302 was summed to obtain the total damage for the entire canton. This procedure of random sampling was repeated
303 1'000 times giving a total damage range for each windstorm event. Unless otherwise stated, for each windstorm
304 the median of the damage distribution is given hereinafter.

305 **2.3.2   CLIMADA impact model**

306 The windstorm damage model in the open source risk assessment platform CLIMADA is relying on open data
307 only and that is why it is deviating in some aspects from GVZ's approach described above. As the windstorm
308 hazard component is open, it is identical to the hazard input used in case of the GVZ damage model. The
309 exposure is based on public data instead of GVZ's proprietary portfolio information. CLIMADA uses produced
310 capital for Switzerland published by the World Bank (2018) as the total value of physical assets for Switzerland
311 and further uses a combination of nightlight intensity and population density to create a reliable geographical
312 distribution of the assets (Eberenz et al., 2019). The resulting values are then distributed to building footprints
313 from OpenStreetMap (OpenStreetMap contributors, 2017). Analogous to the GVZ damage model, CLIMADA
314 uses the MDD curve of Schwierz et al. (2010). Instead of a random resampling of affected buildings, the MDD
315 factor is combined with the deterministic factor "Percentage of Assets Affected" (PAA).

316 As the total value of the exposure is different between the GVZ exposure, the CLIMADA exposure, and the
317 exposure used in Schwierz et al. (2010), the MDD and PAA factors might be wrongly scaled for this study. In
318 the CLIMADA model setup used, we adjusted for this by linearly scaling the MDD and PAA factors to reduce
319 the difference of the modelled damages and the insured damages for matching events (i.e., by minimising the
320 root-mean-square deviation, RMSD). This adjustment conserved the shape of the original vulnerability curve.

The CLIMADA impact model and the GVZ damage model have a different sensitivity to the hazard intensity: in CLIMADA, damage is possible for a wind gust speed of 72 km/h (20 m/s) and above, in the GVZ damage model for 90 km/h (25 m/s) and above.

2.4 Assessment of potential windstorm damage and risk

Risk is defined here as the product of the extent of damage and the probability of damage. The probability of damage is driven, on the one hand, by the probability that the building is within the area of high wind gust speeds and, on the other hand, by the return period of the windstorm event. The probability, that the building is within the area of high wind gust speeds is incorporated in the modelled damage amount by the spatially explicit modelling approach and the vulnerability, which includes the percentage of assets affected (in case of CLIMADA). The return period or frequency of windstorm events is derived from the hazard event sets. Return periods express the probability of occurrence of windstorm events (e.g., an event with a return period of 250 years is expected on average every 250 years).

There are several risk assessment metrics that can be calculated with a set of event damages, which are the main result from the damage modelling described above.

2.4.1 Average annual damage

The average annual damage (AAD) is an important risk measure in the insurance industry. It describes the risk from all events reported on an annual basis:

$$AAD = \frac{sum\ of\ all\ event\ damages}{time\ range\ covered\ by\ event\ set} = \sum_{event\ i} event\ damage_i * annual\ frequeny_i \quad (2)$$

2.4.2 Exceedance frequency curve

Using the annual frequencies of the events in a hazard event set, it is possible to determine at what frequency a certain damage amount is exceeded. The largest damage amount is exceeded once in the time range covered by the damage event set, the second largest damage amount is exceeded twice, the third one thrice and so on. The exceedance frequency curve shows the damage amount as a function of exceedance frequency. For large damage amounts, this matching typically relies on only a few damage events, which increases the statistical uncertainty.

2.4.3 Pareto pricing

In the insurance industry, the concept of "Pareto pricing" is a simple approach to represent and extrapolate the distribution of a damage event set to define the price of insurance contracts (Mitchell-Wallace et al., 2017). We imitated this pricing method by fitting a Generalized Pareto Distribution (GPD) to damage event sets using a Maximum Likelihood Estimate (MLE). We do this even though some assumptions in statistical theory are not valid for these datasets (e.g., windstorm damage event sets are clustered which breaks the independence assumption), as we use the GPD only to show the underlying sampling uncertainty. To fit a GPD to a damage event set, only the threshold has to be chosen. We chose a threshold for each damage event set, which results in a parameterised GPD with similar exceedance frequencies for the largest damage amount in the event set. For

354 the insured damages we chose a threshold of CHF 0.4 million and for the modelled damage event set based on

355 "WISC historic" we chose a threshold of CHF 0.1 million. By using the percent point function (the inverse of a

356 cumulative distribution function) on the fitted distributions, an exceedance frequency curve for the fitted

357 distribution was calculated.

358 To illustrate the statistical uncertainty of the exceedance frequency curve, we undertook a resampling and

359 thereby show the sampling uncertainty for each damage event set. In the resampling, we generated 200 random

360 samples from the fitted distribution and used the MLE to fit a GPD to each random sample. The exceedance

361 frequency curves of these resampled distributions illustrate the uncertainty especially for rare events with a high

362 return period. We show the 90-% confidence interval of damage amounts for each exceedance frequency, which

363 spans from the 5th percentile to the 95th percentile of the 200 samples.

364 In case of the damage event set computed on the basis of "WISC probabilistic extension", the uncertainty is best

365 illustrated by the sampling uncertainty of the damage event set based on "WISC historic" for the following

366 reasons. The procedure of computing the hazard event set "WISC probabilistic extension" by statistical

367 perturbation (as described in Sect. 2.2.3) transforms part of the sampling uncertainty of the hazard event set

368 "WISC historic" into an uncertainty of the parameters α and β in Eq. (1). However, this parameter uncertainty is

369 difficult to illustrate, since no combination of α and β could be found which adequately represents the upper and

370 lower boundaries of the sampling uncertainty of the pan-European SSI distribution. Additionally, the sampling

371 uncertainty of "WISC probabilistic extension" no longer represents the same uncertainty as in case of the other

372 damage event sets. Thus, for comparison of the uncertainties of the different damage event sets, we suggest to

373 use the sampling uncertainty of "WISC historic" as best illustration of the uncertainty of "WISC probabilistic

374 extension".

375 **3   Results**

376 **3.1   Single events**

377 The damage due to Lothar/Martin is by far the largest windstorm event damage in GVZ's insurance claims

378 database (Fig. A3a): Lothar/Martin caused insured damages of CHF 62.4 million. Lothar/Martin is the most

379 damaging windstorm event in the canton of Zurich in both the 34-years period of insurance claims data as well

380 as in the 75-years period of "WISC historic". The damages modelled with the GVZ damage model range

381 between CHF 58.0 million and CHF 69.0 million, and the median of all modelled damages amounts to

382 CHF 62.7 million (Fig. A3b). For Burglind, the most damaging event of the "observed footprints", the modelled

383 damages range between CHF 10.4 million and CHF 14.5 million, with a median of CHF 12.0 million. For

384 comparison, the insured damages amount to CHF 14.2 million. Thus, damages associated with intense

385 windstorm events like Lothar/Martin or Burglind are very well modelled with GVZ's damage modelling

386 approach, providing confidence in the methodology. For all recorded windstorm events since 1981 (including

387 the additional seven windstorms in 2017 and 2018), the RMSD between the insured damage and the median

388 modelled damage amounts to CHF 2.4 million. Furthermore, the example of Burglind shows that our

389 methodology of creating windstorm footprints on the basis of interpolated wind gust observations (Sect. 2.2.4) is

390 suitable for present and probably also for future windstorm events.

3.2 Average annual damage

The average annual damage (AAD) calculated based on the insured damages (i.e., the mean damage over the observational period of 34 years) is almost twice as high as the AAD computed on the basis of "WISC historic" (Table 2). Several factors contribute to the fact that the AAD is higher for the insured damages than for the modelled damages based on "WISC historic": (i) the occurrence of the very intense event Lothar/Martin, along with other intense events, in the relatively short available period of insurance claims data (Fig. A3a), (ii) the higher damages of events in the 5-year return period range (Table 2), and (iii) the different number of events per year considered. The hazard event set "WISC probabilistic extension" was created to best represent the low-frequency tail of the pan-European SSI and not the full distribution of (high frequency) damages in the canton of Zurich. Nevertheless, the modelled AAD based on the GVZ damage model and "WISC probabilistic extension" is close to the AAD of "WISC historic".

3.3 Assessment of risks due to extreme windstorm events

Figure 2 shows GVZ's windstorm risk assessment of building damage, including uncertainty, on the basis of all available data sources. Based on the insurance claims data only, the return period for the extreme windstorm event Lothar/Martin is estimated to be 34 years (blue squares). Based on "WISC historic", the return period for Lothar/Martin is estimated to be 75 years (yellow dots). Based on the hazard event set "WISC probabilistic extension" and using GVZ's approach for damage modelling, the return period for a damage amount due to Lothar/Martin would be around 125 years (red diamonds). These estimates represent the best guess for each damage event set. It is important to note that the quantified uncertainty of the estimate for the return period of Lothar/Martin based on "WISC historic" (yellow ribbon, 25 years to > 500 years) incorporates both the estimate for the insurance claims data (blue ribbon) as well as the estimate based on "WISC probabilistic extension".

The extrapolated event damage with a return period of 250 years amounts to about CHF 500 million for "WISC historic" and using the same method for the insured damages the extrapolated 250-year event damage would be even higher, around CHF 2.4 billion (yellow and blue lines in Fig. 2). Contrary to this, the 250-year event damage amounts to only about CHF 75 million in case of the hazard event set "WISC probabilistic extension" (red diamonds). The 90-% confidence interval, which represents the sampling uncertainty of the extrapolation of the damage exceedance frequency, based on "WISC historic" provides a range for the 250-year return period damage of CHF 19 million to CHF 33 billion (yellow ribbon). As "WISC probabilistic extension" is based on the same historic information this uncertainty also applies to its results. At a return period of 250 years, the quantified uncertainty of the estimate based on "WISC historic" incorporates both the estimate for the insurance claims data as well as the estimate based on "WISC probabilistic extension".

An interesting feature illustrated in Fig. 2 is that at higher return periods the modelled damages on the basis of "WISC probabilistic extension" increase less strongly compared to the two extrapolations based on the fitted distributions. Evident "jumps" in the modelled damage (e.g., at return periods of approximately 30 years, 70 years, and 90 years) result from the discrete categorisation of the individual windstorm events and the assumed degrees of impact, respectively, as applied in GVZ's damage modelling approach (Sect. 2.3.1).

3.4 Reproducibility of the results using CLIMADA

In general, GVZ's proprietary windstorm damage model is suitable for correctly simulating building damage in the canton of Zurich (see Fig. 3, Fig. A3, and Sect. 3.1). Using the calibrated CLIMADA impact model for windstorm damage modelling is also suitable and the corresponding RMSD amounts to CHF 1.5 million for all recorded windstorm events since 1981 for which WISC wind gust footprints are available (excluding the additional windstorms in 2017 and 2018). The statistics in Table 2 calculated using the GVZ damage model were also calculated using the CLIMADA impact model and the results can be found in Table A1. In summary, it can be stated that the setup of the two damage models applied works well and e.g. replicates the order of the events, provides a reasonable modelled damage for historic events (compared to insurance claims data), and both RMSD are sufficiently good.

The exceedance frequency curve of the modelled damages based on "WISC probabilistic extension" and the CLIMADA impact model (green triangles in Fig. 2) show in general lower values compared to the damage modelling using the GVZ approach (red diamonds), in particular for return periods between 30 and 70 years. This difference is also reflected in the scatter plots in Fig. 3, where in Fig. 3a the GVZ damage model shows an overestimation of the damage amount due to the windstorm event Vivian/Wiebke (with insured damage of approximately CHF 11 million), whereas the CLIMADA impact model shows an underestimation for the same event. The reason for this over- and underestimation of the damage in case of events such as Vivian/Wiebke could be due to the hazard or exposure part of the respective model, but is more likely due to the applied vulnerability curve itself. Apparently, the two damage models perform differently for windstorm events in a medium intensity category. This difference between the two models also becomes evident regarding the AAD risk metric: the AAD of the CLIMADA impact model with "WISC historic" amounts to CHF 1.1 million (Table A1) and is thus almost a third smaller than the AAD associated with the GVZ damage model (CHF 1.4 million). In addition, the curve of the modelled damages is much smoother in case of CLIMADA (Fig. 2), which can be explained by the fact that in CLIMADA the smooth curve of the PAA factors is used. This shows the importance of the applied vulnerability curve in the presented damage modelling approach.

3.5 Rapid damage estimation

Rapid damage estimation directly after a windstorm event is very useful for insurance companies to get a first rapid assessment of the damage to be expected and to e.g. assign their staff accordingly. For current windstorm events, the GVZ does this using its damage model and the wind gust footprints based on "observed footprints" (Sect. 2.2.4). The 95th percentile of the wind gust speeds at building level in the affected region of the canton of Zurich, which is also used in GVZ's damage model to categorise windstorm events (Sect. 2.3.1), is used as a rapid indicator of the range of possible damages. This process is illustrated in Fig. 4. With the help of the dataset "WISC probabilistic extension", assessments can also be made about potential damages from unprecedented, extreme windstorm events. The uncertainty of the damage assessment for such extreme events can be visualised by the large number of available (extreme) events. In total, "WISC probabilistic extension" contains 17 events which are potentially more damaging than Lothar/Martin. A (modelled) total damage amount of more than CHF 96 million is associated with the most extreme windstorm event in "WISC probabilistic extension" (Fig. 1). Thus, this windstorm is potentially about 1.5 times as damaging as Lothar/Martin.

465 Figure 4 further shows, by the length of the red bars, the stochastic component in GVZ's damage modelling

466 approach, which tries to approximate the random selection as not every building is equally affected during a

467 windstorm event (Sect. 2.3.1). The range of modelled damages (length of red bars) increases with increasing

468 wind gust speed. On the other hand, the quotient of the range of modelled damages and the median of the

469 damage distribution (red points) generally decreases with increasing wind gust speed. "Jumps" in the modelled

470 damage (e.g., for wind gust speeds lower than 126 km/h) again result from the discrete categorisation of the

471 individual windstorm events in the GVZ damage model.

472 The absolute difference between the modelled damage amount and the corresponding value of the regressed

473 relationship (red points and solid red line in Fig. 4) generally increases with increasing wind gust speed.

474 Accordingly, the number of available wind gust footprints decreases with increasing wind gust speed.

475 **4   Discussion**

476 Any information about the historic risk of winter windstorms in the canton of Zurich contains the record of the

477 event Lothar/Martin. As this is the most damaging event in the record by far, the general risk assessment is

478 connected to the assessment of the return period of such an event damage, which will always be uncertain. We

479 argue that the return period based on the historic windstorm footprints (75 years) is much more reliable than the

480 return period based on the insured damage record (34 years). Well knowing that the two estimates each have

481 overlapping uncertainties, the estimates do not contradict each other. Rather the estimates, as best guesses, can

482 inform varying deterministic risk views. Other information, like the return period of Lothar/Martin's damage

483 amount based on "WISC probabilistic extension" and an independent catalogue of historic windstorms in

484 Switzerland by Stucki et al. (2014) suggest that the return period of such a damage amount could be even rarer

485 than 75 years. This clearly shows the added value that GVZ achieves in its risk assessment through applying the

486 WISC wind data compared to using insurance claims data only − and, above all, through the additional dataset

487 "WISC probabilistic extension". The return period of extreme windstorm events such as Lothar/Martin can now

488 be assessed more reliably.

489 The windstorms Lothar and Martin affected, in addition to Switzerland, in particular France, Belgium,

490 Luxembourg, and Germany. The original industry damage associated with Lothar and Martin amount to

491 approximately EUR 5.8 billion and EUR 2.5 billion, respectively (PERILS, 2020). The return period for

492 exceeding the damage amount due to Lothar alone in all of Europe was estimated to be 15 years by

493 Munich Re (2002) and the return period for the cluster of the three windstorms in December 1999 Anatol

494 (3 December 1999), Lothar, and Martin was estimated to be between 22 and 45 years

495 (Renggli and Zimmerli, 2016). This study shows that it is important to make a distinction between the return

496 period of an event like Lothar/Martin in all of Europe and the return period of this event locally, in a relatively

497 small region. The damage modelling shown in this study, using the event set "WISC historic" and the local

498 exposure information, enables a much more reliable derivation of the return period specific to GVZ than the

499 existing scientific work is able to provide.

500 Based on "WISC historic" and the GVZ damage model, the average annual damage for building damages in the

501 canton of Zurich amounts to CHF 1.4 million according to our calculation and we argue that this is the best

502   available estimate for the AAD. However, this estimation is still uncertain due to the high sampling uncertainty,

503   the uncertainty associated with the assessment of the event Lothar/Martin, and the uncertainty with regard to the

504   damage modelling itself. For comparison, in the last 10 years GVZ has experienced yearly damage from all

505   natural hazards of CHF 16 million and additionally yearly damage by fire of CHF 42 million (all numbers from

506   2018; GVZ annual report, 2018). Compared to the risk from these hazards, the estimated AAD from winter

507   windstorms of CHF 1.4 million is relatively small. However, the occurrence of windstorm events such as

508   Vivian/Wiebke, Lothar/Martin, and Burglind has shown that single windstorms are able to cause huge damage

509   amounts and they are consequently an important causal element when assessing capital requirements.

510   Insurance companies undertake their business under a strict regulatory environment, and having enough capital

511   to cover rare events is one of the regulatory requirements. The damage amount reached on average every

512   250 years is an often-mentioned indicator for such a rare event. However, the insured damages and also the

513   modelled damages based on "WISC historic" do not span a long enough period by far to make an empirical

514   prediction of a damage amount with a return period of 250 years. All methods of extrapolation from these

515   datasets suffer from the sampling uncertainty (shown as confidence intervals in Fig. 2). The hazard event set

516   "WISC probabilistic extension" uses the distribution of pan-European SSI values to create a set of probable

517   events with higher return periods than "WISC historic". The uncertainty of the return periods of such events

518   however cannot considerably be reduced compared to "WISC historic", because it relies on the same historic

519   information. In future studies, the information from dynamical models, which are run for many model years,

520   would help to further reduce this uncertainty. Despite the uncertainty, it can nevertheless be important to study

521   the sensitivity of the 250-year return period damage to changes in the portfolio (like growth or changed building

522   codes), changes in the deductible or other changes. "WISC probabilistic extension" provides windstorm

523   footprints of events with a return period of 250 years (and more), that allow the modelling of damages with

524   changes in the exposure or the vulnerability.

525   It comes as no surprise that the choice of the vulnerability curve in the damage modelling approach applied

526   strongly influences the results of the damage estimation (e.g., Koks and Haer, 2018), and unsurprisingly no

527   optimal "one-size-fits-all" vulnerability curve exists. Every damage model behaves differently, not least because

528   different vulnerability curves are used and each of the damage models has been calibrated differently. The

529   vulnerability curve of Schwierz et al. (2010) is based on movable property and building damages associated

530   with European winter windstorms. The rather general function does not make a distinction between building

531   types, in contrast to other available functions (e.g., Feuerstein et al., 2011). For a modelling setup with focus on

532   the hazard, the vulnerability curve of Schwierz et al. (2010) is however suitable and was successfully applied in

533   earlier studies (e.g., Stucki et al., 2015; Welker et al., 2016). The function does not require detailed information

534   regarding the values at risk, which is certainly an advantage for such insurance and reinsurance companies that

535   do not have detailed exposure data for their damage modelling. A disadvantage of the used vulnerability curve is

536   that it does not implicitly provide a quantification of the uncertainty as a probabilistic vulnerability curve would

537   (e.g., Heneka et al., 2006; Prahl et al., 2012). The quantification of the uncertainty of exposure and vulnerability

538   information was generally omitted in this study to focus on the comparison of the claims and hazard datasets.

539   But of course, for comparison of the presented risk numbers with other studies the uncertainty of the

540   vulnerability and exposure information play a bigger role. The vulnerability assumed in this study and the

corresponding hazard intensity only considers the maximum gust speeds during an event and not the duration of high wind gusts within a windstorm event, which can however have a major impact on the damage to be expected. Taking the windstorm duration into account (e.g., Etienne and Beniston, 2012) could improve our damage modelling, and it is planned to implement this in a future version of GVZ's damage model. Furthermore, it is not considered that buildings are partially adapted to local wind conditions (e.g., multi-storey buildings or exposed buildings located on mountain tops).

Not every building is equally affected during a windstorm event. To take that into account, in the GVZ damage model a random resampling of affected buildings was applied according to an assumed degree of impact (red bars in Fig. 4). The assumed degree of impact was derived according to the respective severity category of the windstorm. This severity categorisation and the assumed degrees of impact are inevitably relatively rough in GVZ's current model setup, because the assumptions are based on insurance claims data from only a few past windstorm events in the canton of Zurich. With every further windstorm, these assumptions will however become more reliable in the future. In contrast, the deterministic PAA values (Schwierz et al. 2010), as used in the CLIMADA impact model, are much smoother and thus allow a smooth damage modelling (Fig. 2). However, these values are not specific for windstorms in the canton of Zurich and they do not allow a stochastic sampling as in GVZ's damage modelling approach.

The rapid estimate of the damage potential in the event of extreme, unprecedented windstorm events shown in Fig. 4 is just one example of how the WISC data and in particular the additional damage event set "WISC probabilistic extension" can be used for insurance applications. The idea was to be able to make a statement about the damage to be expected simply based on available wind observations in the area of the canton of Zurich. It is always important for insurance companies to be able to give a damage assessment as rapidly as possible after an event, not least when it comes to media inquiries. However, one should keep in mind that the uncertainty shown does not incorporate the full uncertainty of the damage estimate, but rather the uncertainty that results from the random selection as not all buildings are affected equally during a windstorm event. In a future study, it would be interesting to quantify the full uncertainty of the rapid damage estimate.

Not least, the WISC wind data enable insurance companies to evaluate the variability and long-term changes of winter windstorms and their associated damage since 1940. Besides a marked interannual and decadal-scale variability of windstorms in the canton of Zurich, we find a tendency for more intense windstorms since approximately mid of the 1980s (Fig. A3d). One possible reason for this positive trend is that "WISC historic" consists of two "parts" with different databases: until 1979, the ERA-20C reanalysis (Poli et al., 2016) was used for downscaling, followed by the ERA-Interim reanalysis (Dee et al., 2011). Furthermore, a change in the large-scale, atmospheric dynamics has been observed in recent decades, which was conducive to increased winter windstorm activity and intensity in Switzerland (Welker and Martius, 2015). This change was accompanied by an atmospheric circulation pattern resembling a southeastwardly displaced winter North Atlantic Oscillation (NAO) pattern. Which of the two reasons is dominant for the found positive tendency in winter windstorm intensity and associated damages in the canton of Zurich could not be finally clarified in the present study. Furthermore, how winter windstorm activity and intensity in mid-latitude Europe will change in a future warmer climate is still uncertain (Catto et al., 2019).

579 **5 Conclusion**

580 This study is an example of how a regional building insurance company in Switzerland uses the open database
581 of European windstorm event sets provided by WISC in combination with a probabilistic extension for their
582 assessment of potential building damages and risks as a result of extreme winter windstorm events, including an
583 evaluation of the uncertainties. The windstorm event Lothar/Martin in December 1999 is the most damaging
584 event in both the insurance claims data and "WISC historic" (damage of more than CHF 60 million). The
585 average annual damage for building damages in the canton of Zurich is CHF 1.4 million, computed based on
586 "WISC historic" and the GVZ damage model.

587 Both the insurance claims data and the modelled building damages based on "WISC historic" are rather
588 unsuitable for evaluating rare windstorm damage events with return periods considerably exceeding the
589 observational period. The new hazard event set "WISC probabilistic extension" projects a damage amount of
590 approximately CHF 75 million for a return period of 250 years, while the uncertainty for an extrapolation to
591 such return periods is still very large. However, the probabilistic hazard event set allows for testing the
592 sensitivity of the risk to e.g. changes in the insurance portfolio or in the insurance condition (e.g., the deductible)
593 for events of a higher intensity than the observed historic events.

594 Our analysis is implemented in GVZ's proprietary windstorm damage model as well as in the open source risk
595 assessment platform CLIMADA (Bresch and Aznar-Siguan, 2019a). This guarantees scientific reproducibility
596 and offers insurance companies and other societal actors in Switzerland and the rest of Europe the opportunity to
597 apply the shown methodology to their own portfolio with a low entry threshold. This study illustrates how open
598 climatological data and open source damage models can be used to assess windstorm risks in Europe and how
599 this approach complements risk assessments based on proprietary insurance claims data only.

600 There is a growing societal need for physical risk disclosure, not least in the context of the Task Force for
601 Climate-related Financial Disclosure (TCFD; Surminski et al., 2020). The presented methodology, in particular
602 the combination of the WISC hazard data with the open source CLIMADA platform, can be used for such a
603 disclosure report.

604 **Code availability and data availability**

605 The scripts reproducing the main results of the paper and the figures are available under
606 https://github.com/CLIMADA-project/climada_papers. The probabilistic hazard event set "WISC probabilistic
607 extension" for each European country is made available for download under https://doi.org/10.3929/ethz-b-
608 000406567 (Röösli and Bresch, 2020).

609 CLIMADA is openly available at GitHub (https://github.com/CLIMADA-project/climada_python; Bresch and
610 Aznar-Siguan, 2019a) under the GNU GPL license (GNU operating system, 2007). The documentation is hosted
611 on Read the Docs (https://climada-python.readthedocs.io/en/stable/; Aznar-Siguan and Bresch, 2019b) and
612 includes a link to the interactive tutorial of CLIMADA. CLIMADA v1.4.1 was used for this publication, which
613 is permanently available at the ETH Data Archive: http://doi.org/10.5905/ethz-1007-252 (Bresch et al., 2020).

Author contribution

CW and TR share first co-authorship and contributed equally to defining the case study, performing the analyses, writing the article, and participating in the review process. CW developed the GVZ damage model and TR generated the hazard event set "WISC probabilistic extension". DNB contributed to writing the article, conceptualised CLIMADA, and over-saw its implementation in Python, based on the previous MATLAB implementation by himself.

Competing interests

The authors declare that they have no conflict of interest.

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786 **Table 1:** Summary statistics for the windstorm hazard event sets and insurance claims data used in this study.

Dataset	Available years (period)	Total number of available windstorm hazard events	Number of damage events in the canton of Zurich
"WISC historic"	75 (1940-2014)	142	27
"WISC probabilistic extension"	2'250 (30*75)	142 (parent events) and 4'118 (altered offspring events)	754
"WISC synthetic"	405 (3*135)	22'980	42
"WISC operational"	39 (1979-2017)	106	untested
"Observed footprints"	2 (2017-2018)	7	7
Insurance claims data	36 (1981-2014 and 2017-2018)	-	18 ("WISC historic") and 7 ("observed footprints")

787 **Table 2:** Annual average damage (AAD) and event damage for different return periods (RP) and the windstorm
788 event Lothar/Martin on the basis of insurance claims data and modelled damages using the GVZ damage model
789 and the hazard event sets "WISC historic" and "WISC probabilistic extension", respectively.

	Available years (period)	AAD [CHF m.]	Event damage with 5-year RP [CHF m.]	Event damage with 10-year RP [CHF m.]	Event damage with 50-year RP [CHF m.]	Event damage with 250-year RP [CHF m.]	Event damage due to Lothar/ Martin [CHF m.]
Insurance claims data	34 (1981-2014)	2.3	0.6	1.1	-	-	62.4
"WISC historic"	75 (1940-2014)	1.4	0.2	1.3	31.4	-	62.7
"WISC probabilistic extension"	2'250 (30*75)	1.4	0.2	1.3	17.0	74.6	-

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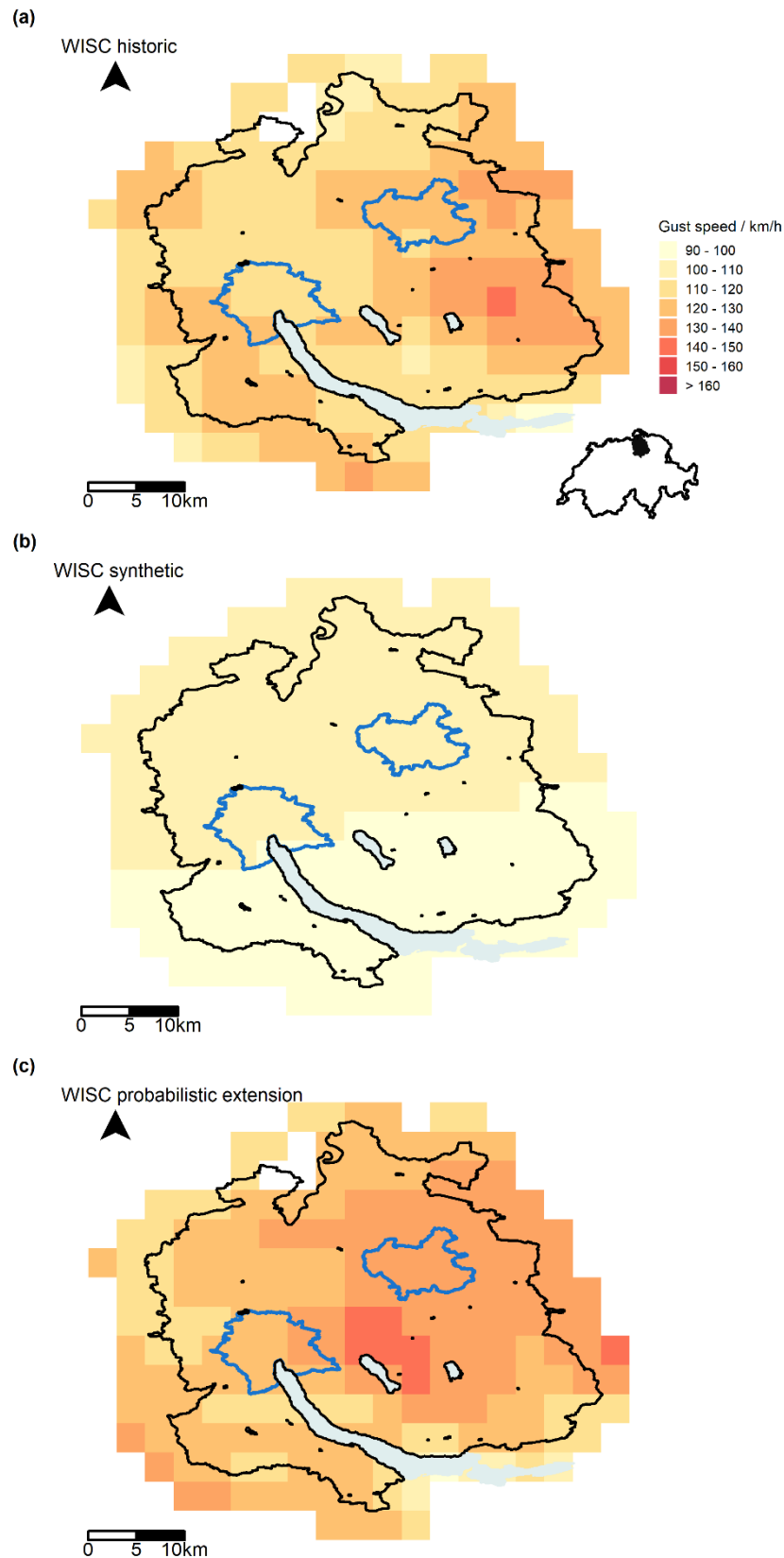
Figure 1: Maximum wind gusts for every grid cell in the canton of Zurich (i.e., windstorm footprints) for the most damaging events in **(a)** "WISC historic", **(b)** "WISC synthetic", and **(c)** "WISC probabilistic extension". The urban areas of the two main cities Zurich (left) and Winterthur (right) are marked in blue.
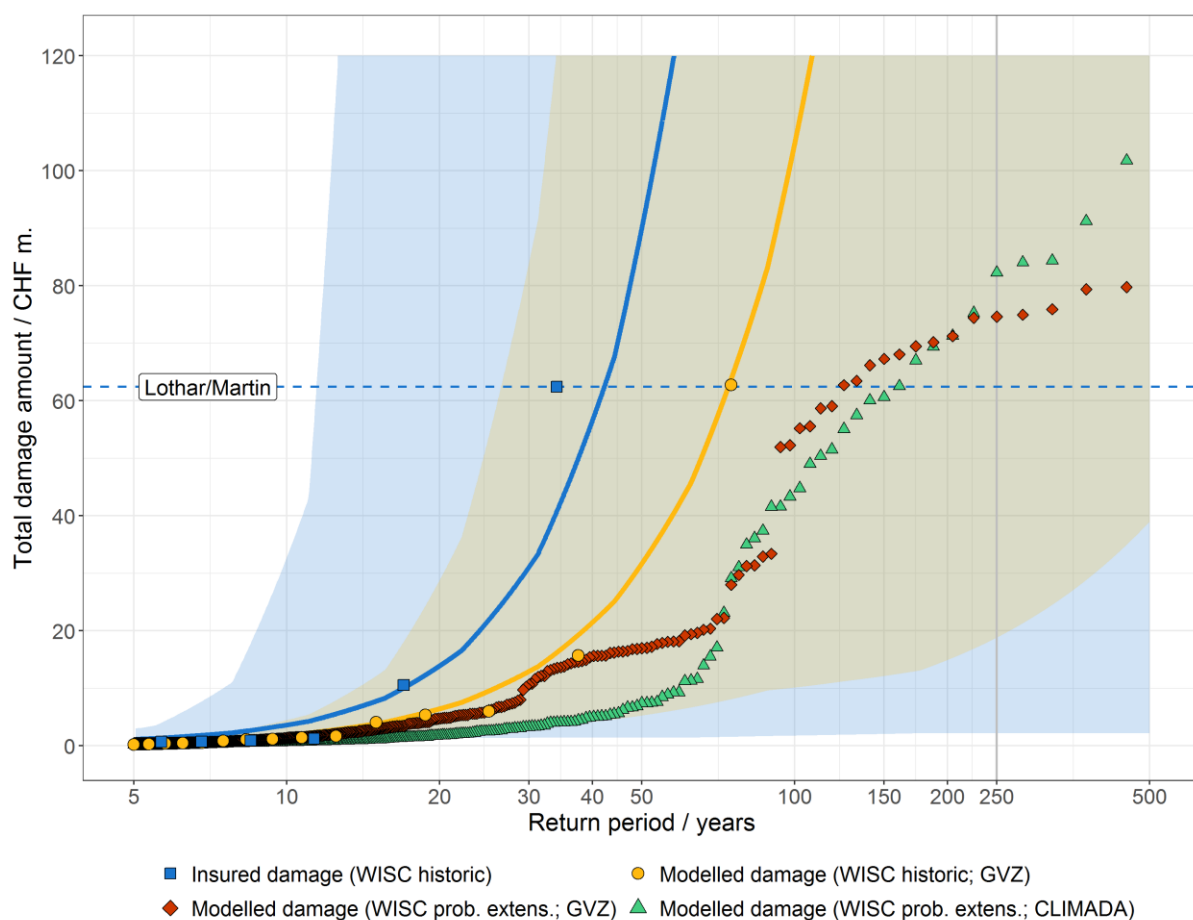
Figure 2: Exceedance frequency curves for building damages in the canton of Zurich based on different data sources. The blue squares indicate the insured damages according to GVZ's database (excluding the additional windstorms in 2017 and 2018), the blue solid line represents a GPD fitted to the insured damages, and the blue ribbon is the 90-% confidence interval produced by resampling. The yellow dots, solid line, and ribbon are analogous to the blue, but for the modelled damages based on "WISC historic" and the GVZ damage model. The red diamonds (green triangles) show the exceedance frequency curve of the modelled damages based on the hazard event set "WISC probabilistic extension" and the GVZ damage model (CLIMADA). The insured total damage for Lothar/Martin is shown by a blue dashed horizontal line, and the 250-year return period is indicated by a grey solid vertical line.
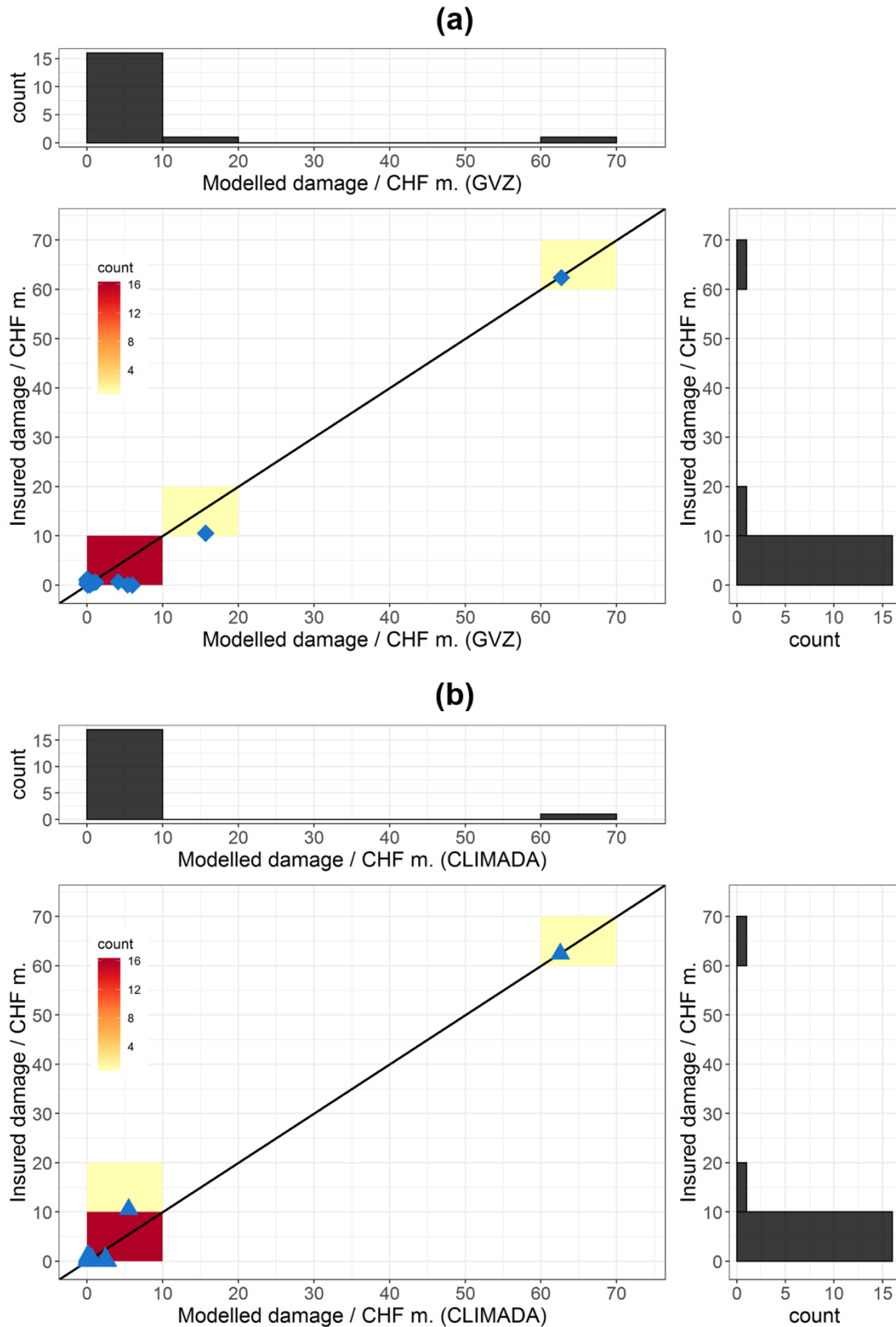
Figure 3: 2d-histograms for the normalised insured total damages in the canton of Zurich versus the modelled total damages based on **(a)** the GVZ damage model (diamonds) and **(b)** the CLIMADA impact model (triangles), respectively, for all windstorms with damage > 0 in the hazard event set "WISC historic". Marginal histograms are shown in the top and right panels.

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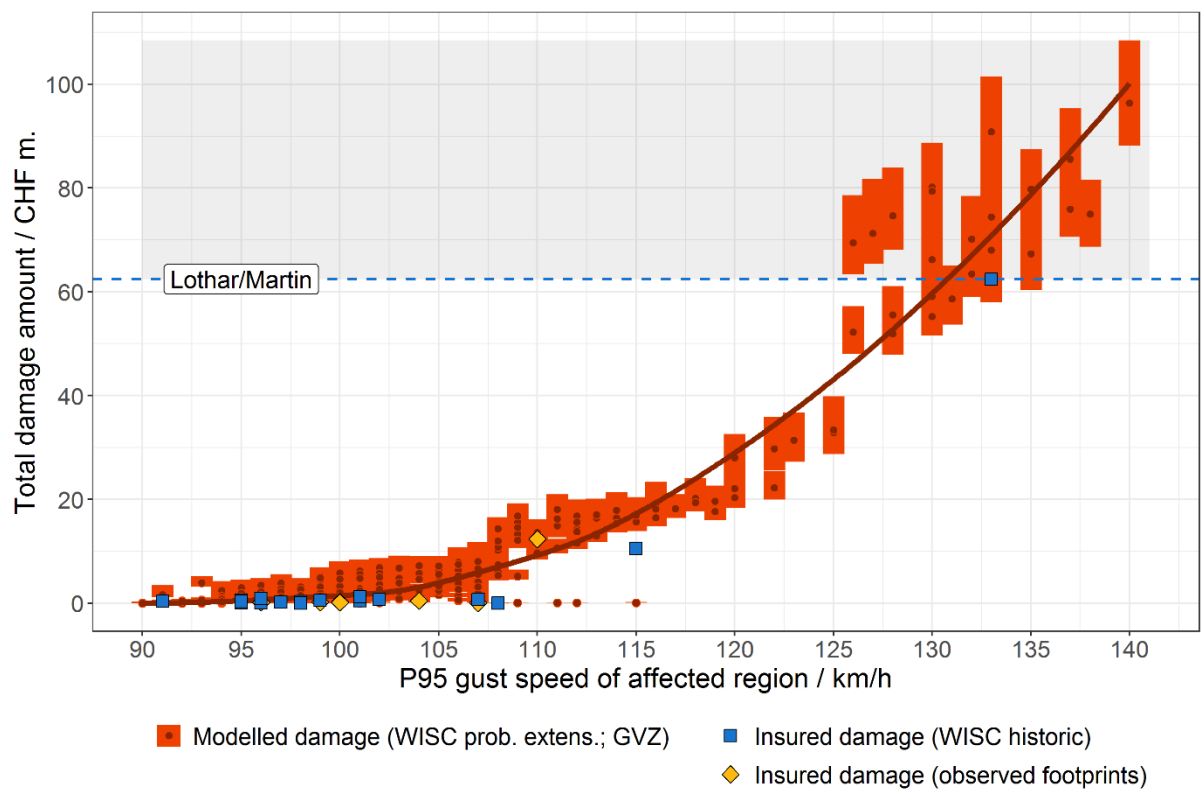
Figure 4: Total damage modelled using the GVZ damage model and the hazard event set "WISC probabilistic extension" versus the 95th percentile of the corresponding gust speeds in the affected region of the canton of Zurich (median of 1'000 random damage modelling as red points; range of modelled damages indicated as red bars). The 95th percentile of the gust speeds is shown, because the 95th percentile is used in GVZ's damage model to categorise windstorm events (Sect. 2.3.1). The relationship between wind gust speed and modelled total damage is further approximated by a locally estimated scatterplot smoothing (LOESS) and a bootstrap method (i.e., random resampling with replacement, number of samples = 1'000; median of confidence interval given as solid red line). Furthermore, the relationship between gust speeds and normalised insured total damages based on "WISC historic" and independent, interpolated wind gust observations (selection of windstorms in 2017 and 2018, including winter windstorm Burglind) are given as blue squares and yellow diamonds, respectively. The domain for unprecedented windstorm damages − i.e. beyond Lothar/Martin − is shaded grey.

 Appendix

824 **Table A1:** AAD and event damage for different return periods (RP) and the windstorm event Lothar/Martin on
825 the basis of insurance claims data and modelled damages using the CLIMADA impact model and the hazard
826 event sets "WISC historic" and "WISC probabilistic extension", respectively.

	Available years (period)	AAD [CHF m.]	Event damage with 5-year RP [CHF m.]	Event damage with 10-year RP [CHF m.]	Event damage with 50-year RP [CHF m.]	Event damage with 250-year RP [CHF m.]	Event damage due to Lothar/ Martin [CHF m.]
Insurance claims data	34 (1981-2014)	2.3	0.6	1.1	-	-	62.4
"WISC historic"	75 (1940-2014)	1.1	0.2	0.6	24.5	-	62.6
"WISC probabilistic extension"	2'250 (30*75)	1.2	0.2	0.6	7.4	82.3	-

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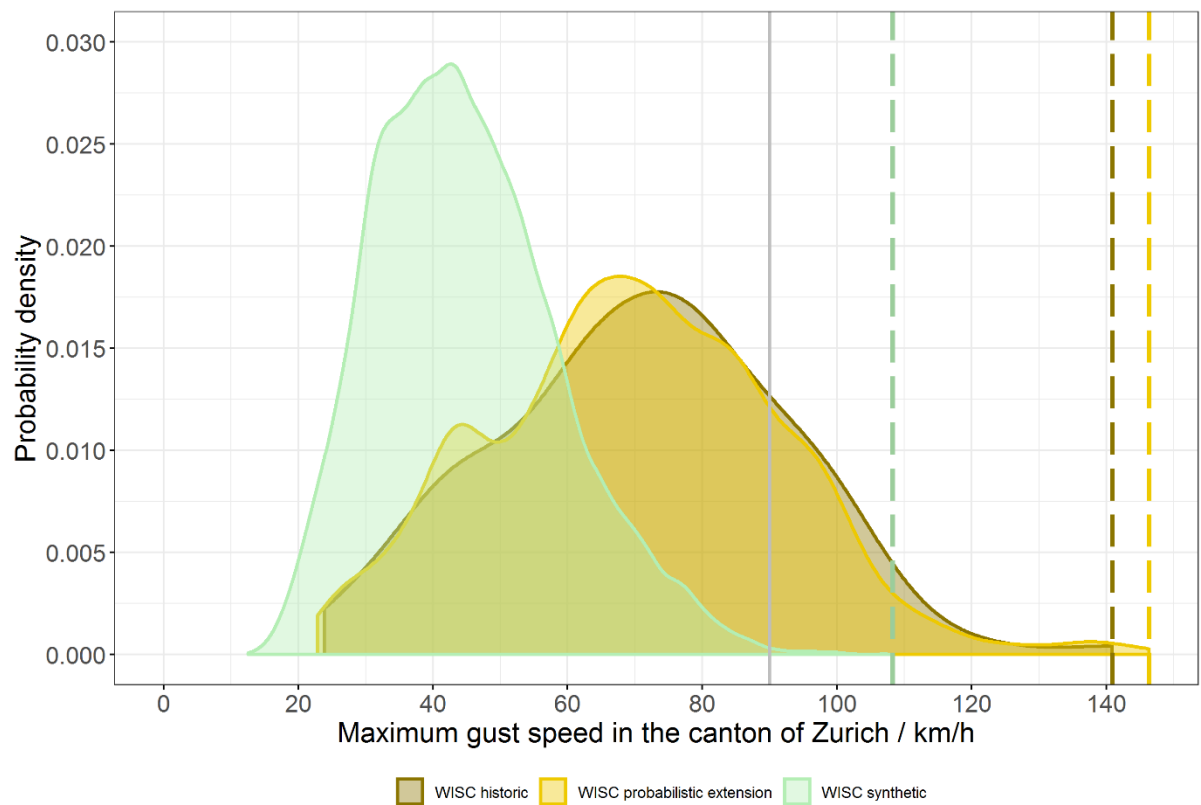
Figure A1: Probability density functions of the maximum gust speeds at building level in the canton of Zurich
for the three hazard event sets "WISC historic" (brown), "WISC probabilistic extension" excluding the parent
windstorms (yellow), and "WISC synthetic" (green). The maxima of the individual distributions are shown as
dashed vertical lines. In the GVZ damage model, damage is possible from a wind gust speed of more than
90 km/h, which is here indicated by a grey solid vertical line.

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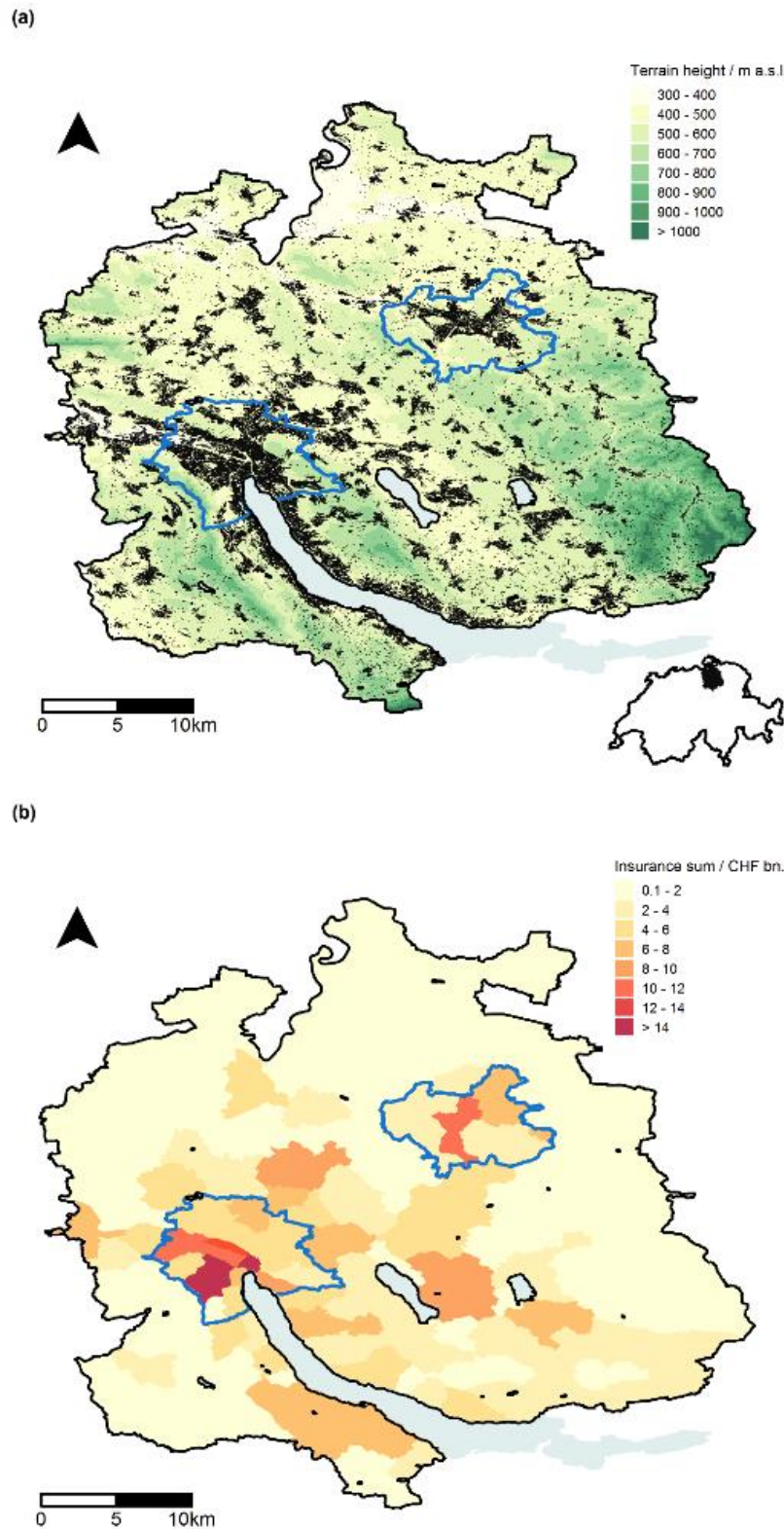
Figure A2: (a) Terrain height for the canton of Zurich (colour scheme) according to a digital elevation model with a horizontal grid size of 200 m (source: Swiss Federal Office of Topography; Swisstopo DEM, 2019). In addition, the spatial distribution of all buildings insured by GVZ is indicated and the urban areas of the two main cities, Zurich (left) and Winterthur (right), are marked in blue. **(b)** Total building sum insured for each municipality (colour scheme).

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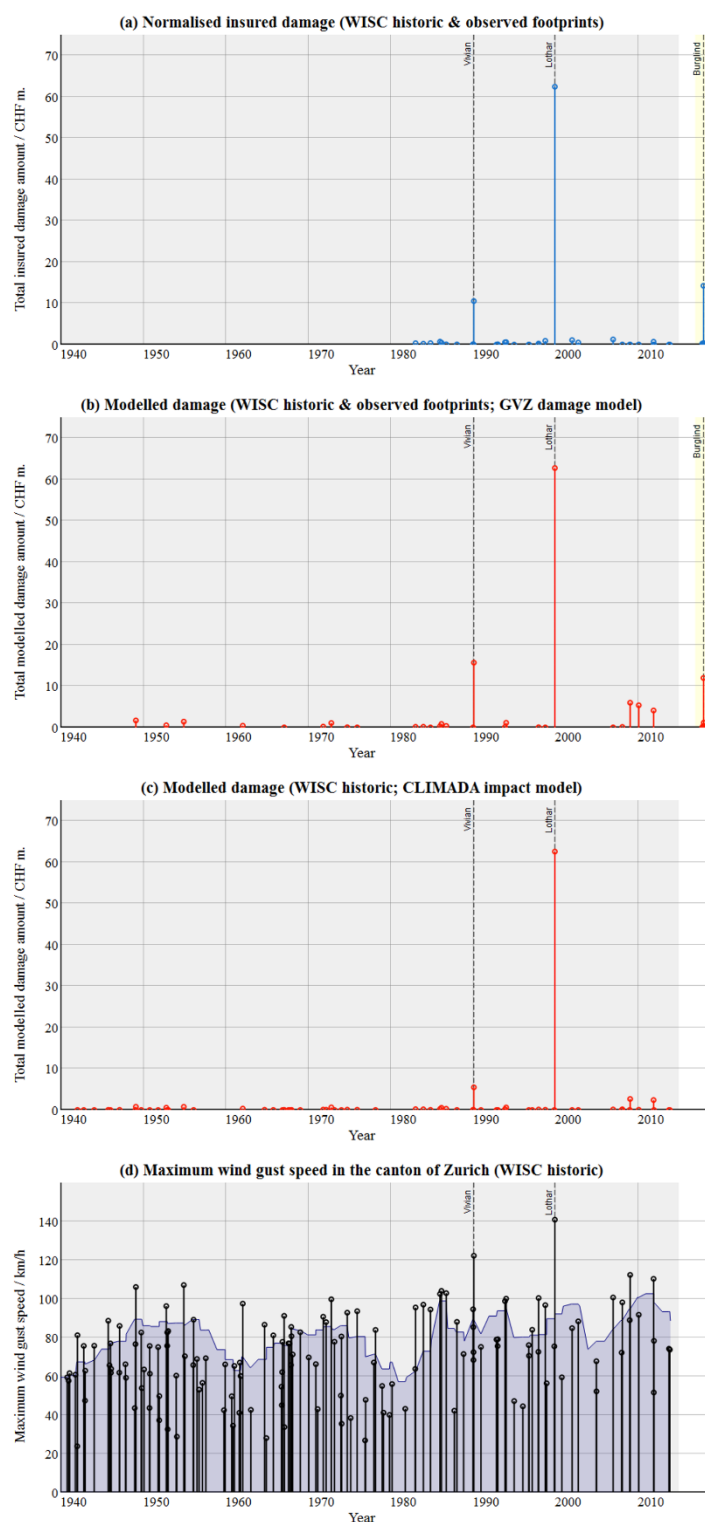
Figure A3: Variability of windstorms and associated damages in the canton of Zurich: **(a)** normalised insured damage, **(b)** modelled windstorm damage based on the GVZ damage model and the hazard event sets "WISC historic" and "observed footprints", **(c)** modelled windstorm damage based on the CLIMADA impact model and "WISC historic", and **(d)** maximum gust speeds at building level in the canton of Zurich according to "WISC historic" (black stem plot). The filled time series in (d) additionally shows the 5-year moving average of the yearly maximum gust speeds in the canton of Zurich. The period for which "WISC historic" hazard data ("observed footprints") is available is shaded grey (yellow) in (a) and (b). The windstorm events Vivian/Wiebke, Lothar/Martin, and Burglind are marked.