



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: the 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. Our study  
21 emphasises the importance of complementing a claims-based perspective with a probabilistic risk modelling  
22 approach to better understand windstorm risks. The presented open source model provides a straightforward  
23 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. A larger sample of events is  
58 needed for which at least quantitative meteorological data and if possible damage data at ideally high  
59 spatiotemporal resolution are available (e.g., Haas and Pinto, 2012). Observational damage data are generally  
60 sparse and incomplete for historic windstorms in Switzerland (Stucki et al., 2014). Instead, societal actors often  
61 use modelled impacts to manage their risk. Insurance and reinsurance companies apply impact models for their  
62 pricing and governments use modelled risk for option appraisal (e.g., The Economics of Climate Adaptation  
63 Working Group, 2009; Bresch, 2016). Additionally, the information is needed for climate-related financial  
64 disclosure (Surminski et al., 2020). However, only very few impact models are available open source and free  
65 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; Prahel 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 ADAPtion; 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 data basis  
98 available to most insurance companies. The windstorm hazard event sets as provided by WISC form an  
99 independent database to validate and further develop existing European winter windstorm models. The dataset  
100 can be used for both pan-European analyses and local analyses, as shown in this study.

101 Using the WISC historic hazard event set allows GVZ in a way to “re-check” historic events. By means of the  
102 synthetic hazard event set, the tail of the hazard and damage distributions should be investigated. However,  
103 Röögli et al. (2018) found that the synthetic hazard event set is not suitable for this purpose. Therefore, we  
104 propose instead a probabilistic windstorm hazard event set based on a method described in  
105 Schwierz et al. (2010) to overcome the shortcomings of the WISC synthetic hazard event set. This new  
106 probabilistic hazard event set of around 4'300 events contains windstorms from the WISC historic hazard event  
107 set altered by various perturbations.

108 This study shows how GVZ uses both the WISC dataset and the new probabilistic hazard event set for assessing  
109 the potential building damage and risk due to extreme windstorm events (including an evaluation of the  
110 uncertainties of such assessments). A relationship between wind gust speed in the entire area of the canton of  
111 Zurich and associated building damages is found, which allows for a rapid, straightforward estimation of  
112 damage directly after the occurrence of extreme, unprecedented windstorms. This study further shows how GVZ



113 was able to improve its windstorm risk assessment on the basis of the WISC dataset and the new probabilistic  
114 hazard event set, and could serve as an example for other players in the insurance sector or other societal actors  
115 in Switzerland and in the rest of Europe. At the same time, this study also illustrates selected limitations of the  
116 WISC dataset.

## 117 2 Data and methods

118 After a description of the insurance claims data (Sect. 2.1) and the windstorm hazard event sets used (Sect. 2.2),  
119 we introduce the GVZ and the CLIMADA risk assessment models applied for damage modelling (Sect. 2.3) and  
120 conclude this section with a brief recapitulation of the risk assessment metrics employed in this study  
121 (Sect. 2.4).

### 122 2.1 Insurance claims data

123 The windstorm damages of past events are recorded in a proprietary database of GVZ. It consists of almost  
124 40 years of insurance claims data, in total more than 84'000 single wind damage records. From this database all  
125 the events relevant for this study were selected by following the event definition of the windstorm event set  
126 “WISC historic” (Sect. 2.2.1). In total, 18 events are associated with WISC windstorms based on that definition  
127 (see also Table 1). Due to the nature of the database, only the damage reports actually insured by GVZ were  
128 considered. The insurance claims data allow GVZ to assess the risk for its own portfolio by analysing frequency  
129 and severity of past damages, i.e. to assess its risk due to winter windstorm events with a return period smaller  
130 than 40 years. Additional information can help GVZ to put their recorded damages into reference and to get a  
131 better estimate of the risk of events with a return period larger than the 40 years of experience.

132 For the sake of comparability, the insured damages had to be normalised to present-day exposure levels. In this  
133 study, the applied normalisation considers the general inflation on the basis of the Zurich construction price  
134 index (2020). Hereinafter, both insured and modelled windstorm damages are including occasional deductibles  
135 – so-called “gross damages”, to ease comparison.

### 136 2.2 Windstorm hazard event sets

137 Atmospheric models provide information about winter windstorm events that can be used as hazard component  
138 in a risk assessment model. WISC published several hazard datasets each containing a set of windstorm events  
139 and providing the maximum wind gust per geographic location per event. We used the historic windstorm  
140 footprints (Sect. 2.2.1) and constructed a probabilistic extension based on it (Sect. 2.2.3). In addition, we derived  
141 wind gust footprints from measurements for a selection of present windstorm events (Sect. 2.2.4). The additional  
142 windstorm hazard event sets published by WISC, that are however not considered in this study, are briefly  
143 summarised in Sect. 2.2.2.

#### 144 2.2.1 Historic windstorm hazard event set

145 The historic windstorm hazard event set – denoted “WISC historic” – contains wind gust footprints for around  
146 140 winter windstorm events in Europe in 1940-2014 (i.e., 75 modelled years in total). The events were  
147 selected, on the one hand, based on the high damage they caused and, on the other hand, because of their high



148 intensity in meteorological terms (i.e., high vorticity). Because of this pan-European perspective, the dataset is  
149 not necessarily specific to windstorms in the canton of Zurich. Nevertheless, the high-impact windstorms  
150 Lothar/Martin (26–28 December 1999) and other intense windstorms such as Vivian/Wiebke (26 February–  
151 1 March 1990) are included.

152 The windstorm footprints were computed by running the UK Met Office Unified Model (MetUM; Davies et al.,  
153 2005) at approximately 4.4 km resolution with ERA-20C reanalysis (Poli et al., 2016) and ERA-Interim  
154 reanalysis (Dee et al., 2011) as boundary conditions, covering Europe and parts of the North Atlantic. ERA-20C  
155 was used for all windstorm events in 1940-1979 and ERA-Interim for all events in 1979-2014.

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

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

### 178 **2.2.2 Other WISC hazard event sets**

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

181 1. The operational windstorm hazard event set – denoted “WISC operational” – contains around  
182 110 windstorm events in 1979-2017 and thus more recent events than the windstorm hazard event set “WISC  
183 historic” used in this study, which contains windstorm events until 2014 only. “WISC operational” is based on a  
184 new generation of atmospheric reanalysis, the ERA5 reanalysis (Hersbach and Dee, 2016). As it does not cover



185 the time range 1940-1979 (compared to “WISC historic”) it does not complement the recorded damages by  
186 providing information about historic events not covered by GVZ’s claims database.

187 2. The synthetic windstorm hazard event set – denoted “WISC synthetic” – was created within the  
188 UPSCALE (UK on PRACE - weather-resolving Simulations of Climate for globAL Environmental risk;  
189 UPSCALE, 2020) modelling framework and is a physically realistic set of plausible winter windstorm events in  
190 the period 1985-2011 based on the climatic conditions of that period. The modelling framework developed five  
191 ensembles. The dataset contains wind gust footprints for around 23’000 synthetic windstorms: i.e., three sets of  
192 7’660 events each. Each of the three sets covers 135 modelled years. The original idea of the hazard event set  
193 “WISC synthetic” was to use wind information from climate models to provide wind gust footprints for winter  
194 windstorms in Europe with a return period of 250 years or even higher. However, this hazard event set was not  
195 considered because the findings of Rööslı et al. (2018) could be replicated in this study, showing that the dataset  
196 does not contain the maximum wind gust speeds we would expect from the distribution of the historic  
197 windstorm hazard events (Fig. A1) nor the high intensities we would expect from very rare, high-impact  
198 windstorm events (Fig. 1).

199 For a detailed description of all unused windstorm hazard events sets provided by WISC, we refer to the  
200 documentations available online at WISC products (2019) and WISC hazard event set description (2019).

### 201 2.2.3 Probabilistic windstorm hazard extension

202 Based on “WISC historic”, we generated an additional probabilistic windstorm hazard event set – denoted  
203 “WISC probabilistic extension”. By applying a method described in Schwierz et al. (2010), the individual  
204 windstorm events in “WISC historic” (parent events) were altered to create 29 altered offspring events by  
205 various perturbations: e.g., spatial displacement and by weakening / intensifying the wind speeds (non-altered  
206 wind speeds are spatially displaced only). The spatial displacement was undertaken by shifting the respective  
207 windstorm footprint by about 20 km to the north, south, west, or east. The wind gust speeds were intensified and  
208 weakened by no more than 3 m/s (normally much less) according to the probabilistic alteration of wind speeds  
209 in Eq. (1), with a scale parameter  $\alpha = 0.0225$  and a power parameter  $\beta = 1.15$ :

$$\begin{aligned} \text{windspeed}_{\text{scenario } 1} &= \text{windspeed}_{\text{original}} + \alpha * \text{windspeed}_{\text{original}}^{\beta} \\ \text{windspeed}_{\text{scenario } 2} &= \text{windspeed}_{\text{original}} - \alpha * \text{windspeed}_{\text{original}}^{\beta} \\ \text{windspeed}_{\text{scenario } 3} &= \text{windspeed}_{\text{original}} + \alpha * \sqrt{\beta} \sqrt{\text{windspeed}_{\text{original}}} \\ \text{windspeed}_{\text{scenario } 4} &= \text{windspeed}_{\text{original}} - \alpha * \sqrt{\beta} \sqrt{\text{windspeed}_{\text{original}}} \\ \text{windspeed}_{\text{scenario } 5} &= \text{windspeed}_{\text{original}} + \frac{\alpha}{2} * \text{windspeed}_{\text{original}}^{\beta} \\ &\quad + \frac{\alpha}{2} * \sqrt{\beta} \sqrt{\text{windspeed}_{\text{original}}} \end{aligned} \tag{1}$$



210 These newly created “probabilistic” footprints can be viewed as scenarios of plausible windstorms as they only  
211 differ slightly from historic events, retaining both the spatial extent and general structure.

212 For using the scenarios in a qualitative risk assessment framework, the probabilistic windstorm footprints can be  
213 used as they are, but for a quantitative risk assessment the frequencies of the windstorm footprints need to be  
214 estimated. In an effort to assign reasonable frequency estimates to the probabilistic windstorm footprints, we  
215 considered the distribution of the historic, pan-European Storm Severity Index (SSI; e.g., Lamb and Frydendahl,  
216 1991; Leckebusch et al., 2008; Dawkins et al., 2016). Similar as in Schwierz et al. (2010), the algorithm of  
217 creating the probabilistic windstorm footprints was configured to recreate the cumulative distribution function of  
218 a generalized extreme value (GEV) distribution fitted to the historic SSI values. We defined the frequency of all  
219 probabilistic windstorm footprints to be equal and to sum up to the frequency of the parent windstorm. We then  
220 selected a set of parameters for weakening and intensifying the wind speeds (parameters  $\alpha$  and  $\beta$  in Eq. (1)) that  
221 resulted in a similar probabilistic distribution of SSI as the extrapolated distribution from the historic SSI values.  
222 For the probabilistic hazard event set to best represent the tail of the historic distribution, we determined a  
223 combination of  $\alpha$  and  $\beta$ , that minimises the difference in the cumulative distribution functions for events that are  
224 rarer than 75 years.

225 “WISC probabilistic extension” includes footprints for 4’118 probabilistic windstorm events, along with the  
226 142 original windstorm events in “WISC historic” (Table 1), and provides a basis of an event-based risk  
227 assessment for winter windstorms with return periods of around 250 years, a scenario relevant for regulatory  
228 requirements in the insurance sector. It is important to note that this method incorporates a lot of uncertainty,  
229 including but not limited to the sampling uncertainty of rare events in a relatively short time range (i.e., 75 years  
230 in case of “WISC historic”).

231 Encouragingly, the hazard event set “WISC probabilistic extension” shows considerably higher wind gust  
232 speeds in the canton of Zurich as compared with “WISC synthetic” (Fig. 1). Nonetheless, the maximum wind  
233 gust speeds of the most extreme event in “WISC probabilistic extension” are not considerably higher than those  
234 of Lothar/Martin, the most extreme event in both “WISC historic” and the insurance claims data.

#### 235 **2.2.4 Observed footprints for current windstorms**

236 Real-time wind gust observations can serve as the hazard part of the damage model for a rapid damage  
237 estimation directly after the occurrence of an extreme windstorm event. Such “observed” windstorm footprints  
238 can also be used for further validation of GVZ’s damage modelling approach (Sect. 2.3). To create such  
239 footprints, we used interpolated wind gust measurements in the canton of Zurich based on the Common  
240 Information Platform for Natural Hazards (GIN; GIN platform, 2019) for a selection of seven winter  
241 windstorms in the years 2017 and 2018. With the exception of winter windstorm Burglind hitting Switzerland  
242 on 3 January 2018, the windstorms considered caused only minor damages in the canton of Zurich. The  
243 individual windstorm footprints are based on a total of around 110 measurement stations in the canton of Zurich  
244 and in the immediate vicinity (i.e., buffer zone with a distance of 20 km around the polyline of the canton). For  
245 spatial interpolation, we applied an Inverse Distance Weighting (IDW) interpolation with the Shepard method  
246 used for weight calculation. In this study, the gridded wind gust footprints derived from measurements have a  
247 horizontal resolution of 2 km. The topography of the canton of Zurich is not considered in the applied



248 interpolation method and unquestionably the quality of the derived windstorm footprints could be improved by  
249 using a more elaborate interpolation method, which takes account of the topography.

### 250 **2.3 Damage modelling approaches**

251 The windstorm footprints of the different hazard event sets described in the previous section were used as input  
252 for damage modelling and GVZ's proprietary windstorm damage model was applied for this (Sect. 2.3.1). In  
253 addition, the CLIMADA impact model was used to be able to publish the method used in this study with open  
254 data and open source code (Sect. 2.3.2).

255 In both damage models, the extent of damage results from the intensity of the windstorm event (i.e., hazard), the  
256 value of the asset (i.e., exposure), and the susceptibility of the asset to damage (i.e., vulnerability). This concept  
257 is broadly used and is explained in more detail in Aznar-Siguan and Bresch (2019a). In this study, the  
258 windstorm hazard assessment is based on the winter windstorm footprints described in Sect. 2.2. The exposure  
259 is the value of the buildings in the canton of Zurich and the vulnerability is described by a functional  
260 relationship that defines how much the buildings are damaged at a certain wind gust speed. In both damage  
261 models, we use the vulnerability curve of Schwierz et al. (2010). This vulnerability curve combines the damage  
262 degree and the percentage of assets affected. Only damage to buildings is estimated. The estimate does not  
263 include damage to movable property, damage to infrastructure, nor business interruption.

#### 264 **2.3.1 GVZ damage model**

265 The damage estimates in this model are computed using a rather conventional modelling framework and the  
266 reduced complexity of the approach allows a well interpretable assessment of the model skill. Normally, GVZ  
267 uses its damage model directly after the occurrence of a windstorm event to estimate the expected building  
268 damage. Furthermore, GVZ applies the damage model to estimate the damage potential and the risk associated  
269 with windstorms with regard to solvency considerations and prevention options. The main points of the  
270 modelling approach are described in the following.

271 The initial step is a simple spatial overlay of the gridded maximum wind gust speeds during the respective  
272 windstorm event with GVZ's current building stock (from 2018; without sublevel garages, as they are usually  
273 not affected by windstorms), where GVZ's proprietary building database with information about e.g. the sum  
274 insured of each building and the publicly available building footprints (GIS browser Zurich, 2019) were used.  
275 GVZ's insurance penetration in the canton of Zurich is almost 100 %. In the damage model, damage is possible  
276 from a wind gust speed of more than 90 km/h, and only buildings affected by such gusts were considered in the  
277 following modelling steps.

278 Figure A2 shows the spatial distribution of all insured buildings in the canton of Zurich as well as of the total  
279 sum insured at municipal level. The aggregated sum insured for all buildings in the two main cities, Zurich and  
280 Winterthur (municipal boundaries indicated by blue polygons), accounts for almost 40 % of the total insured  
281 value for the entire canton.

282 To estimate the damage in monetary terms, the value of each individual building (i.e., its insured value) was  
283 multiplied by the factor "Mean Damage Degree" (MDD) according to Schwierz et al. (2010), where the gust





284 speeds at building level computed in the first step were converted into the corresponding MDD factors. The  
285 MDD factors are a non-linear function of the maximum wind gust speed during a windstorm event. The same  
286 vulnerability curve of Schwierz et al. (2010) is also implemented in the open source impact model CLIMADA  
287 (Aznar-Siguan and Bresch, 2019a). The vulnerability curve is diagrammed in Welker et al. (2016).

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

### 299 **2.3.2 CLIMADA impact model**

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

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

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

### 318 **2.4 Assessment of potential windstorm damage and risk**

319 Risk is defined here as the product of the extent of damage and the probability of damage. The probability of  
320 damage is driven, on the one hand, by the probability that the building is within the area of high wind gust



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

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

#### 329 **2.4.1 Average annual damage**

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

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

#### 332 **2.4.2 Exceedance frequency curve**

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

#### 339 **2.4.3 Pareto pricing**

340 In the insurance industry, the concept of “Pareto pricing” is a simple approach to represent and extrapolate the  
341 distribution of a damage event set to define the price of insurance contracts (Mitchell-Wallace et al., 2017). We  
342 imitated this pricing method by fitting a Generalized Pareto Distribution (GPD) to damage event sets using a  
343 Maximum Likelihood Estimate (MLE). We do this even though some assumptions in statistical theory are not  
344 valid for these datasets (e.g., windstorm damage event sets are clustered which breaks the independence  
345 assumption), as we use the GPD only to show the underlying sampling uncertainty. To fit a GPD to a damage  
346 event set, only the threshold has to be chosen. We chose a threshold for each damage event set, which results in  
347 a parameterised GPD with similar exceedance frequencies for the largest damage amount in the event set. For  
348 the insured damages we chose a threshold of CHF 0.4 million and for the modelled damage event set based on  
349 “WISC historic” we chose a threshold of CHF 0.1 million. By using the percent point function (the inverse of a  
350 cumulative distribution function) on the fitted distributions, an exceedance frequency curve for the fitted  
351 distribution was calculated.

352 To illustrate the statistical uncertainty of the exceedance frequency curve, we undertook a resampling. In the  
353 resampling, we generated 200 random samples from the fitted distribution and used the MLE to fit a GPD to  
354 each random sample. The exceedance frequency curves of these resampled distributions illustrate the



355 uncertainty especially for rare events with a high return period. We show the 90-% confidence interval of  
356 damage amounts for each exceedance frequency, which spans from the 5th percentile to the 95th percentile of  
357 the 200 samples.

### 358 **3 Results**

#### 359 **3.1 Single events**

360 The damage due to Lothar/Martin is by far the largest windstorm event damage in GVZ's insurance claims  
361 database (Fig. A3a): Lothar/Martin caused insured damages of CHF 62.4 million. Lothar/Martin is the most  
362 damaging windstorm event in the canton of Zurich in both the 34-years period of insurance claims data as well  
363 as in the 75-years period of "WISC historic". The damages modelled with the GVZ damage model range  
364 between CHF 58.0 million and CHF 69.0 million, and the median of all modelled damages amounts to  
365 CHF 62.7 million (Fig. A3b). For Burglind, the most damaging event of the "observed footprints", the modelled  
366 damages range between CHF 10.4 million and CHF 14.5 million, with a median of CHF 12.0 million. For  
367 comparison, the insured damages amount to CHF 14.2 million. Thus, damages associated with intense  
368 windstorm events like Lothar/Martin or Burglind are very well modelled with GVZ's damage modelling  
369 approach, providing confidence in the methodology. For all recorded windstorm events since 1981 (including  
370 the additional seven windstorms in 2017 and 2018), the RMSD between the insured damage and the median  
371 modelled damage amounts to CHF 2.4 million. Furthermore, the example of Burglind shows that our  
372 methodology of creating windstorm footprints on the basis of interpolated wind gust observations (Sect. 2.2.4) is  
373 suitable for present and probably also for future windstorm events.

#### 374 **3.2 Average annual damage**

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

#### 385 **3.3 Assessment of risks due to extreme windstorm events**

386 Figure 2 shows GVZ's windstorm risk assessment of building damage, including uncertainty, on the basis of all  
387 available data sources. Based on the insurance claims data only, the return period for the extreme windstorm  
388 event Lothar/Martin is estimated to be 34 years (blue squares). Based on "WISC historic", the return period for  
389 Lothar/Martin is estimated to be 75 years (yellow dots). Based on the hazard event set "WISC probabilistic



390 extension” and using GVZ’s approach for damage modelling, the return period for a damage amount due to  
391 Lothar/Martin would be around 125 years (red diamonds).

392 The extrapolated event damage with a return period of 250 years amounts to about CHF 500 million for  
393 “WISC historic” and using the same method for the insured damages the extrapolated 250-year event damage  
394 would be even higher, around CHF 2.4 billion (yellow and blue lines in Fig. 2). Contrary to this, the 250-year  
395 event damage amounts to only about CHF 75 million in case of the hazard event set “WISC probabilistic  
396 extension” (red diamonds). The 90-% confidence interval, which represents the sampling uncertainty of the  
397 extrapolation of the damage exceedance frequency, based on “WISC historic” provides a range for the 250-year  
398 return period damage of CHF 19 million to CHF 33 billion (yellow ribbon). As “WISC probabilistic extension”  
399 is based on the same historic information this uncertainty also applies to its results.

400 Interesting to see in Fig. 2 is that the tail of the modelled damages on the basis of “WISC probabilistic  
401 extension” is reaching far smaller damages per return period than the two extrapolations based on the fitted  
402 distributions. Evident “jumps” in the modelled damage (e.g., at return periods of approximately 30 years,  
403 70 years, and 90 years) result from the discrete categorisation of the individual windstorm events and the  
404 assumed degrees of impact, respectively, as applied in GVZ’s damage modelling approach (Sect. 2.3.1).

#### 405 **3.4 Reproducibility of the results using CLIMADA**

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

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



428 (Fig. 2), which can be explained by the fact that in CLIMADA the smooth curve of the PAA factors is used.  
429 This shows the importance of the applied vulnerability curve in the presented damage modelling approach.

### 430 3.5 Rapid damage estimation

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

443 Figure 4 further shows, by the length of the red bars, the stochastic component in GVZ’s damage modelling  
444 approach, which tries to approximate the random selection as not every building is equally affected during a  
445 windstorm event (Sect. 2.3.1). The range of modelled damages (length of red bars) increases with increasing  
446 wind gust speed. On the other hand, the quotient of the range of modelled damages and the median of the  
447 damage distribution (red points) generally decreases with increasing wind gust speed. “Jumps” in the modelled  
448 damage (e.g., for wind gust speeds lower than 126 km/h) again result from the discrete categorisation of the  
449 individual windstorm events in the GVZ damage model.

450 The absolute difference between the modelled damage amount and the corresponding value of the regressed  
451 relationship (red points and solid red line in Fig. 4) generally increases with increasing wind gust speed.  
452 Accordingly, the number of available wind gust footprints decreases with increasing wind gust speed.

## 453 4 Discussion

454 Any information about the historic risk of winter windstorms in the canton of Zurich contains the record of the  
455 event Lothar/Martin. As this is the most damaging event in the record by far, the general risk assessment is  
456 connected to the assessment of the return period of such an event damage, which will always be uncertain. We  
457 argue that the return period based on the historic windstorm footprints (75 years) is much more reliable than the  
458 return period based on the insured damage record (34 years). Other information, like the return period of  
459 Lothar/Martin’s damage amount based on “WISC probabilistic extension” and an independent catalogue of  
460 historic windstorms in Switzerland by Stucki et al. (2014) suggest that the return period of such a damage  
461 amount could be even rarer than 75 years. This clearly shows the added value that GVZ achieves in its risk  
462 assessment through applying the WISC wind data compared to using insurance claims data only – and, above  
463 all, through the additional dataset “WISC probabilistic extension”. The return period of extreme windstorm  
464 events such as Lothar/Martin can now be assessed more reliably.



465 The windstorms Lothar and Martin affected, in addition to Switzerland, in particular France, Belgium,  
466 Luxembourg, and Germany. The original industry damage associated with Lothar and Martin amount to  
467 approximately EUR 5.8 billion and EUR 2.5 billion, respectively (PERILS, 2020). The return period for  
468 exceeding the damage amount due to Lothar alone in all of Europe was estimated to be 15 years by  
469 Munich Re (2002) and the return period for the cluster of the three windstorms in December 1999 Anatol  
470 (3 December 1999), Lothar, and Martin was estimated to be between 22 and 45 years  
471 (Renggli and Zimmerli, 2016). This study shows that it is important to make a distinction between the return  
472 period of an event like Lothar/Martin in all of Europe and the return period of this event locally, in a relatively  
473 small region. The damage modelling shown in this study, using the event set “WISC historic” and the local  
474 exposure information, enables a much more reliable derivation of the return period specific to GVZ than the  
475 existing scientific work is able to provide.

476 Based on “WISC historic” and the GVZ damage model, the average annual damage for building damages in the  
477 canton of Zurich amounts to CHF 1.4 million according to our calculation and we argue that this is the best  
478 available estimate for the AAD. However, this estimation is still uncertain due to the high sampling uncertainty,  
479 the uncertainty associated with the assessment of the event Lothar/Martin, and the uncertainty with regard to the  
480 damage modelling itself. For comparison, in the last 10 years GVZ has experienced yearly damage from all  
481 natural hazards of CHF 16 million and additionally yearly damage by fire of CHF 42 million (all numbers from  
482 2018; GVZ annual report, 2018). Compared to the risk from these hazards, the estimated AAD from winter  
483 windstorms of CHF 1.4 million is relatively small. However, the occurrence of windstorm events such as  
484 Vivian/Wiebke, Lothar/Martin, and Burglind has shown that single windstorms are able to cause huge damage  
485 amounts and they are consequently an important causal element when assessing capital requirements.

486 Insurance companies undertake their business under a strict regulatory environment, and having enough capital  
487 to cover rare events is one of the regulatory requirements. The damage amount reached on average every  
488 250 years is an often-mentioned indicator for such a rare event. However, the insured damages and also the  
489 modelled damages based on “WISC historic” do not span a long enough period by far to make an empirical  
490 prediction of a damage amount with a return period of 250 years. All methods of extrapolation from these  
491 datasets suffer from the sampling uncertainty (shown as confidence intervals in Fig. 2). The hazard event set  
492 “WISC probabilistic extension” uses the distribution of pan-European SSI values to create a set of probable  
493 events with higher return periods than “WISC historic”. The uncertainty of the return periods of such events  
494 however cannot considerably be reduced compared to “WISC historic”, because it relies on the same historic  
495 information. Despite the uncertainty, it can nevertheless be important to study the sensitivity of the 250-year  
496 return period damage to changes in the portfolio (like growth or changed building codes), changes in the  
497 deductible or other changes. “WISC probabilistic extension” provides windstorm footprints of events with a  
498 return period of 250 years (and more), that allow the modelling of damages with changes in the exposure or the  
499 vulnerability.

500 It comes as no surprise that the choice of the vulnerability curve in the damage modelling approach applied  
501 strongly influences the results of the damage estimation (e.g., Koks and Haer, 2018), and unsurprisingly no  
502 optimal “one-size-fits-all” vulnerability curve exists. Every damage model behaves differently, not least because  
503 different vulnerability curves are used and each of the damage models has been calibrated differently. The



504 vulnerability curve of Schwierz et al. (2010) is based on movable property and building damages associated  
505 with European winter windstorms. The rather general function does not make a distinction between building  
506 types, in contrast to other available functions (e.g., Feuerstein et al., 2011). For a modelling setup with focus on  
507 the hazard, the vulnerability curve of Schwierz et al. (2010) is however suitable and was successfully applied in  
508 earlier studies (e.g., Stucki et al., 2015; Welker et al., 2016). The function does not require detailed information  
509 regarding the values at risk, which is certainly an advantage for such insurance and reinsurance companies that  
510 do not have detailed exposure data for their damage modelling. The vulnerability assumed in this study and the  
511 corresponding hazard intensity only considers the maximum gust speeds during an event and not the duration of  
512 high wind gusts within a windstorm event, which can however have a major impact on the damage to be  
513 expected. Taking the windstorm duration into account (e.g., Etienne and Beniston, 2012) could improve our  
514 damage modelling, and it is planned to implement this in a future version of GVZ's damage model.  
515 Furthermore, it is not considered that buildings are partially adapted to local wind conditions (e.g., multi-storey  
516 buildings or exposed buildings located on mountain tops).

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

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

536 Not least, the WISC wind data enable insurance companies to evaluate the variability and long-term changes of  
537 winter windstorms and their associated damage since 1940. Besides a marked interannual and decadal-scale  
538 variability of windstorms in the canton of Zurich, we find a tendency for more intense windstorms since  
539 approximately mid of the 1980s (Fig. A3d). One possible reason for this positive trend is that "WISC historic"  
540 consists of two "parts" with different databases: until 1979, the ERA-20C reanalysis (Poli et al., 2016) was used  
541 for downscaling, followed by the ERA-Interim reanalysis (Dee et al., 2011). Furthermore, a change in the large-  
542 scale, atmospheric dynamics has been observed in recent decades, which was conducive to increased winter



543 windstorm activity and intensity in Switzerland (Welker and Martius, 2015). This change was accompanied by  
544 an atmospheric circulation pattern resembling a southeastwardly displaced winter North Atlantic Oscillation  
545 (NAO) pattern. Which of the two reasons is dominant for the found positive tendency in winter windstorm  
546 intensity and associated damages in the canton of Zurich could not be finally clarified in the present study.  
547 Furthermore, how winter windstorm activity and intensity in mid-latitude Europe will change in a future warmer  
548 climate is still uncertain (Catto et al., 2019).

## 549 **5 Conclusion**

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

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

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

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

## 574 **Code availability and data availability**

575 The scripts reproducing the main results of the paper and the figures is available under  
576 [https://github.com/CLIMADA-project/climada\\_papers](https://github.com/CLIMADA-project/climada_papers). The probabilistic hazard event set “WISC probabilistic





577 extension” for each European country is made available for download under <https://doi.org/10.3929/ethz-b->  
578 000406567 (Röösli and Bresch, 2020).

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

584

#### 585 **Author contribution**

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

#### 591 **Competing interests**

592 The authors declare that they have no conflict of interest.

#### 593 **Acknowledgements**

594 We are very thankful to the WISC consortium and project team for making all the data and documentation  
595 available and fully open access. Map data copyrighted OpenStreetMap contributors and available from  
596 <https://www.openstreetmap.org>. We want to thank Jan Hartman for his help implementing the Storm Europe  
597 hazard module in the Python version of CLIMADA, Evelyn Mühlhofer for implementing the OpenStreetMap  
598 exposure module in CLIMADA, as well as Samuel Eberenz and Maurice Skelton for providing valuable input  
599 on the manuscript.

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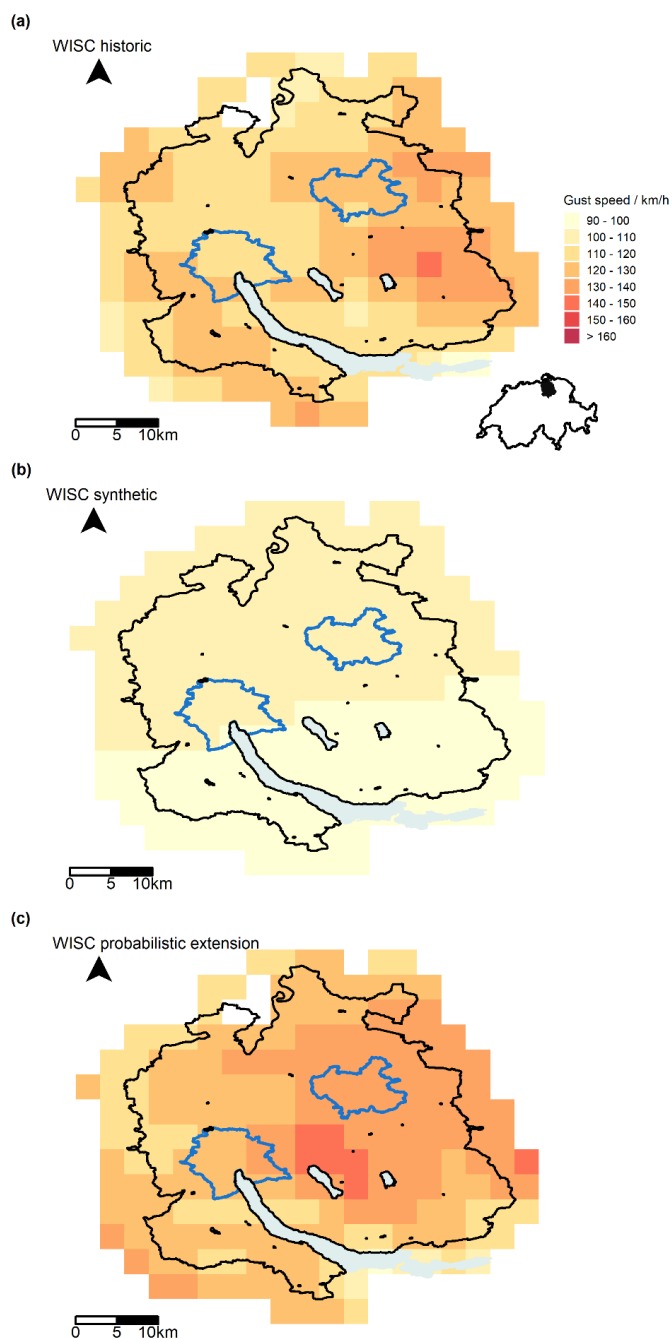
751 **Table 1:** Summary statistics for the windstorm hazard event sets and insurance claims data used in this study.

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”)

752 **Table 2:** Annual average damage (AAD) and event damage for different return periods (RP) and the windstorm  
 753 event Lothar/Martin on the basis of insurance claims data and modelled damages using the GVZ damage model  
 754 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	-

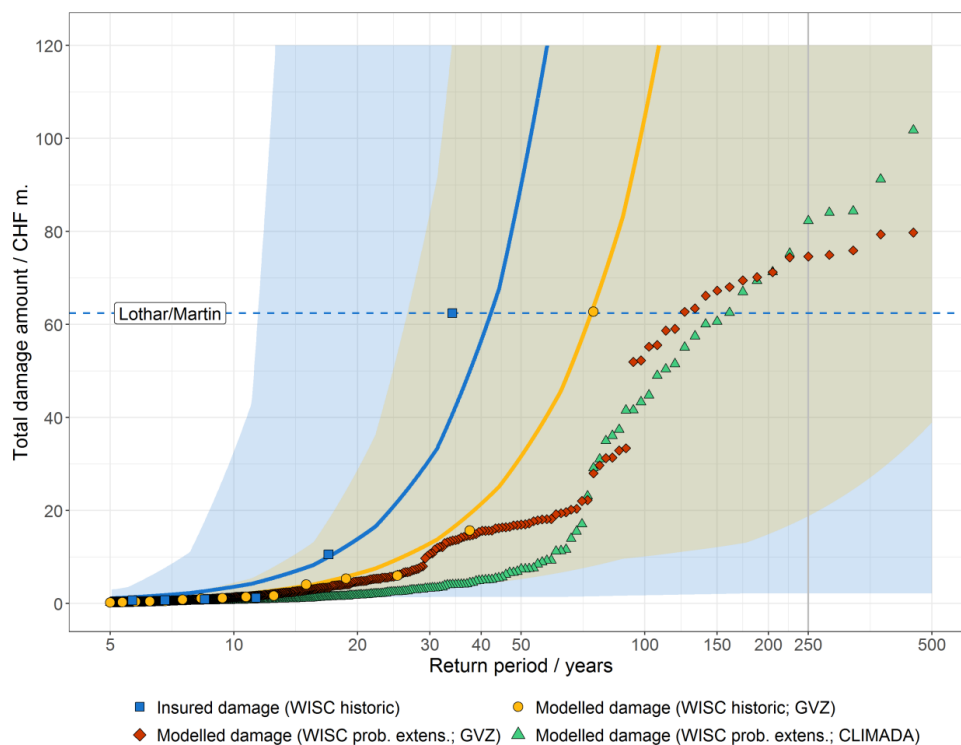
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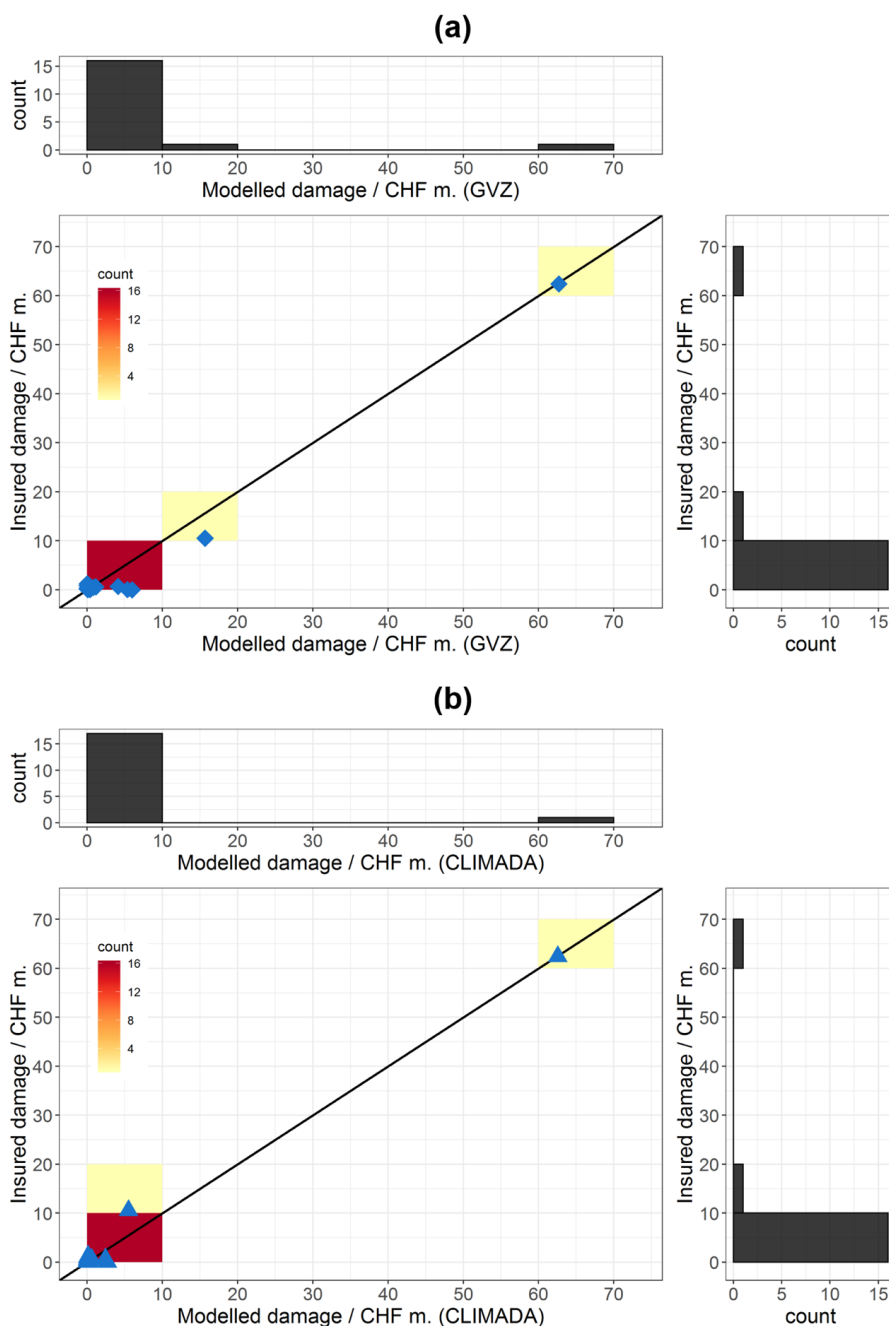
757 **Figure 1:** Maximum wind gusts for every grid cell in the canton of Zurich (i.e., windstorm footprints) for the  
758 most damaging events in (a) “WISC historic”, (b) “WISC synthetic”, and (c) “WISC probabilistic extension”.  
759 The urban areas of the two main cities Zurich (left) and Winterthur (right) are marked in blue.





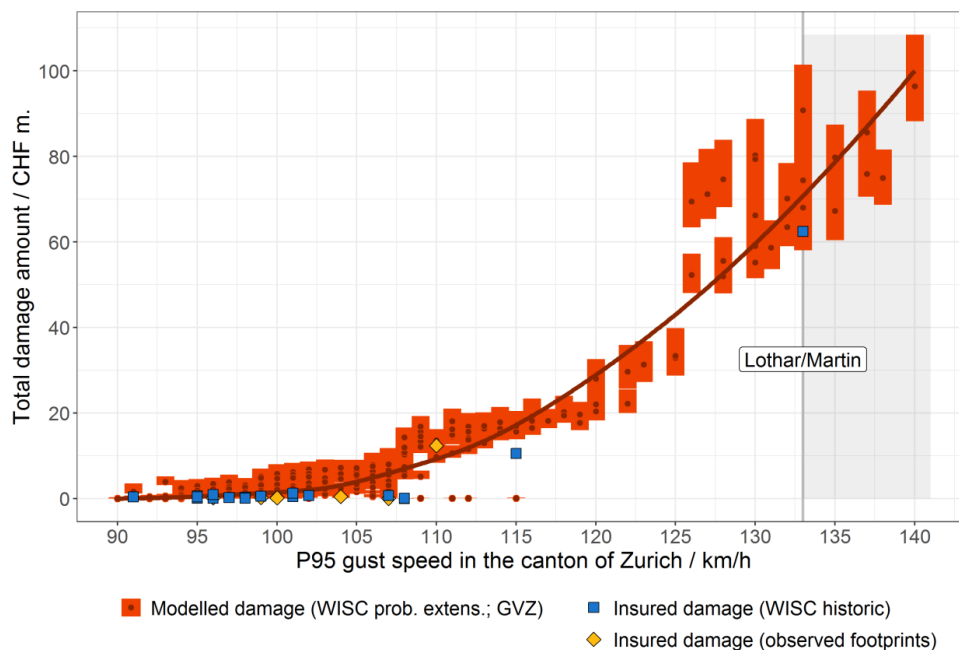
760

761 **Figure 2:** Exceedance frequency curves for building damages in the canton of Zurich based on different data  
762 sources. The blue squares indicate the insured damages according to GVZ's database (excluding the additional  
763 windstorms in 2017 and 2018), the blue solid line represents a GPD fitted to the insured damages, and the blue  
764 ribbon is the 90-% confidence interval produced by resampling. The yellow dots, solid line, and ribbon are  
765 analogous to the blue, but for the modelled damages based on "WISC historic" and the GVZ damage model.  
766 The red diamonds (green triangles) show the exceedance frequency curve of the modelled damages based on the  
767 hazard event set "WISC probabilistic extension" and the GVZ damage model (CLIMADA). The insured total  
768 damage for Lothar/Martin is shown by a blue dashed horizontal line, and the 250-year return period is indicated  
769 by a grey solid vertical line.



770

771 **Figure 3:** 2d-histograms for the normalised insured total damages in the canton of Zurich versus the modelled  
 772 total damages based on **(a)** the GVZ damage model (diamonds) and **(b)** the CLIMADA impact model  
 773 (triangles), respectively, for all windstorms with damage > 0 in the hazard event set “WISC historic”. Marginal  
 774 histograms are shown in the top and right panels.



775

776 **Figure 4:** Total damage modelled using the GVZ damage model and the hazard event set “WISC probabilistic  
777 extension” versus the 95th percentile of the corresponding gust speeds in the canton of Zurich (median of  
778 1’000 random damage modelling as red points; range of modelled damages indicated as red bars). The 95th  
779 percentile of the gust speeds is shown, because the 95th percentile is used in GVZ’s damage model to categorise  
780 windstorm events (Sect. 2.3.1). The relationship between wind gust speed and modelled total damage is further  
781 approximated by a locally estimated scatterplot smoothing (LOESS) and a bootstrap method (i.e., random  
782 resampling with replacement, number of samples = 1’000; median of confidence interval given as solid red  
783 line). Furthermore, the relationship between gust speeds and normalised insured total damages based on “WISC  
784 historic” and independent, interpolated wind gust observations (selection of windstorms in 2017 and 2018,  
785 including winter windstorm Burglind) are given as blue squares and yellow diamonds, respectively. The domain  
786 for unprecedented windstorms – i.e. beyond Lothar/Martin – is shaded grey.

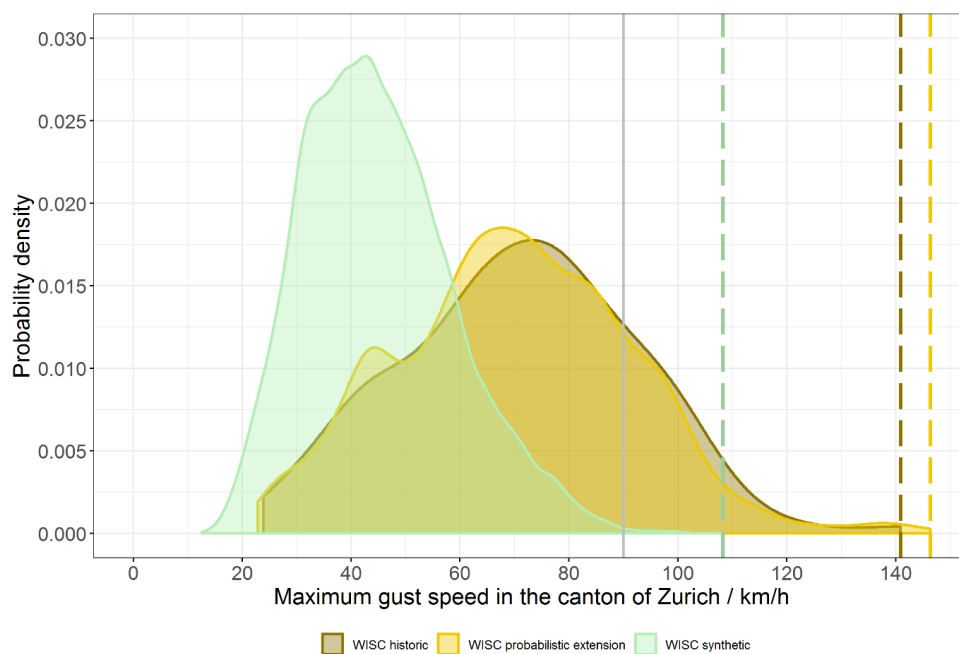


787 **Appendix**

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

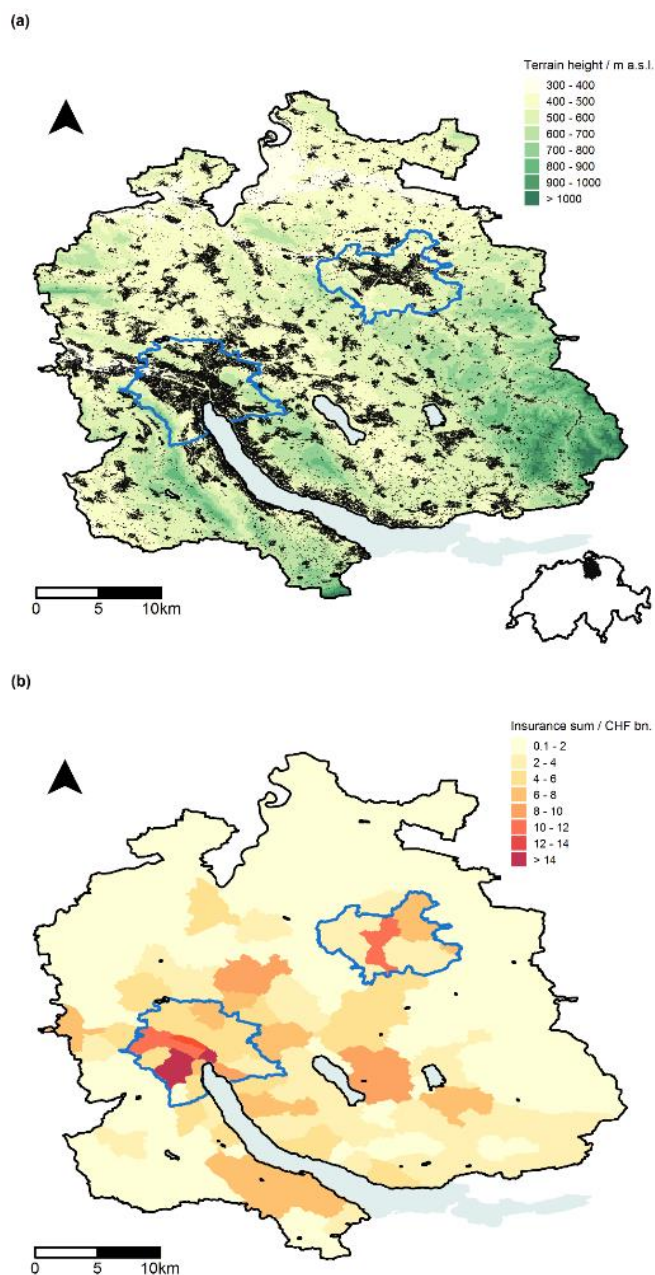
	Available years (period)	AAD [CHF m.]	Event damage with 5- year RP [CHF m.]	Event damage with 10- year RP [CHF m.]	Event damage with 50- year RP [CHF m.]	Event damage with 250- year RP [CHF m.]	Event damage due to Lothar/ Martin [CHF m.]
Insurance claims data	34 (1981- 2014)	2.3	0.6	1.1	-	-	62.4
“WISC 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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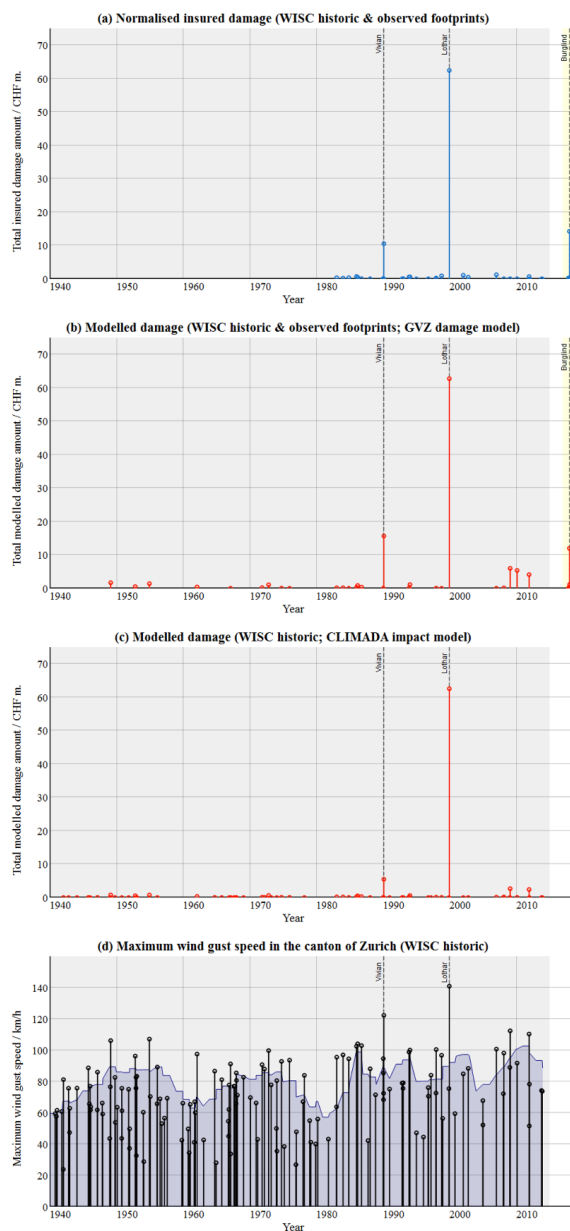
792

793 **Figure A1:** Probability density functions of the maximum gust speeds at building level in the canton of Zurich  
794 for the three hazard event sets “WISC historic” (brown), “WISC probabilistic extension” excluding the parent  
795 windstorms (yellow), and “WISC synthetic” (green). The maxima of the individual distributions are shown as  
796 dashed vertical lines. In the GVZ damage model, damage is possible from a wind gust speed of more than  
797 90 km/h, which is here indicated by a grey solid vertical line.



798

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



804

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