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C susy.mdpi.com/user/review/review/43681770/bm2Qkgwq G 🖻 🕁 dio.cessar@yahoo.com My Profile Logout Submit Journals Topics Information Author Services Initiatives About **Review Acceptance** 0 ∨User Menu Home Agriculture (ISSN 2077-0472) Journal Manage Accounts Manuscript ID agriculture-2705964 Change Password Туре Article Edit Profile Title Climatic Relationships to Agricultural Insurance Loss for the Pacific Northwest Region of the United States Logout Erich Seamon * , Paul E. Gessler , John T. Abatzoglou , Philip W. Mote , Stephen S. Lee Authors Section Agricultural Economics, Policies and Rural Management Submissions Menu ค Special Issue Feature Papers in the Fields of Agricultural Economics, Policies and/or Rural Management Submit Manuscript Abstract Agricultural crop insurance is an important component for mitigating farm risk, particularly given the potential Display Submitted for unexpected climatic events. Using a 2.8 million nationwide insurance claim dataset from the United Manuscripts States Department of Agriculture (USDA), this research study examines spatiotemporal variations of over 31,000 agricultural insurance loss claims across the 24-county region of the inland Pacific Northwest English Editing (iPNW) portion of the United States, from 2001 to 2022. Wheat is the dominant insurance loss crop for the Discount Vouchers region, accounting for over 2.8 billion dollars in indemnities, with over 1.5 billion dollars resulting in claims Invoices due to drought. While fruit production generates considerably lesser insurance losses (400 million dollars) as a primary result of freeze, frost and hail, overall revenue ranks number one for the region, with over 2 LaTex Word Count billion dollars in sales. Principal components analysis of crop insurance claims showed distinct spatial and temporal differentiation in wheat and apples insurance losses using the range of damage causes as factor loadings. The first two factor loadings for wheat account for approximately 50 percent of total variance for Ø ~Reviewers Menu the region, with apples having 60 percent variance. Reviews 1 Volunteer Preferences Information Guidelines for https://www.mdpi.com/reviewers Reviewers Instructions for Authors https://www.mdpi.com/journal/agriculture/instructions

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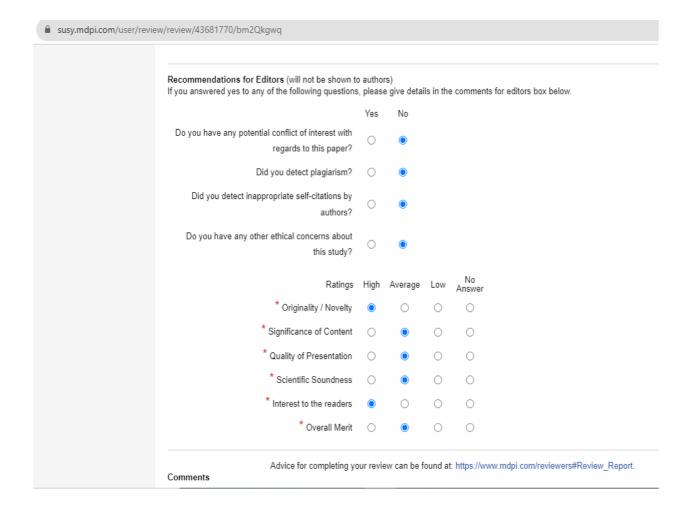
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Home Manage Accounts Change Password Edit Profile Logout Submissions Menu Submitsions Menu Submit Manuscript Display Submitted Manuscripts English Editing Discount Vouchers Invoices LaTex Word Count Reviewers Menu	Journal Manuscript ID Type Title Authors Section Special Issue Abstract	Erich Seamon * Agricultural Ecor Feature Papers i Agricultural crop for unexpected o States Departme 31,000 agricultur (IPNW) portion o region, accounti due to drought. \ as a primary ress billion dollars in : temporal differer loadings. The first	sig64 nships to Agricultural Insurance Loss , Paul E. Gessler , John T. Abatzogi nomics, Policies and Rural Manager in the Fields of Agricultural Econom v insurance is an important compone climatic events. Using a 2.8 million n ent of Agriculture (USDA), this resea ral insurance loss claims across the of the United States, from 2001 to 20 mg for over 2.8 billion dollars in to 10 While fruit production generates con sult of freeze, frost and hail, overall r sales. Principal components analysis ntiation in wheat and apples insuran	for the Pacific Northwest Region of the United Sta bu , Philip W. Mote , Stephen S. Lee nent cs, Policies and/or Rural Management nt for mitigating farm risk, particularly given the pot ationwide insurance claim dataset from the United rch study examines spatiotemporal variations of ov 24-county region of the inland Pacific Northwest 22. Wheat is the dominant insurance loss crop minities, with over 1.5 billion dollars resulting in clair siderably lesser insurance losses (400 million dolli evenue ranks number one for the region, with over s of crop insurance claims showed distinct spatial a ce losses using the range of damage causes as fac unt for approximately 50 percent of total variance for	rential ver the ms ars) 2 and ctor		
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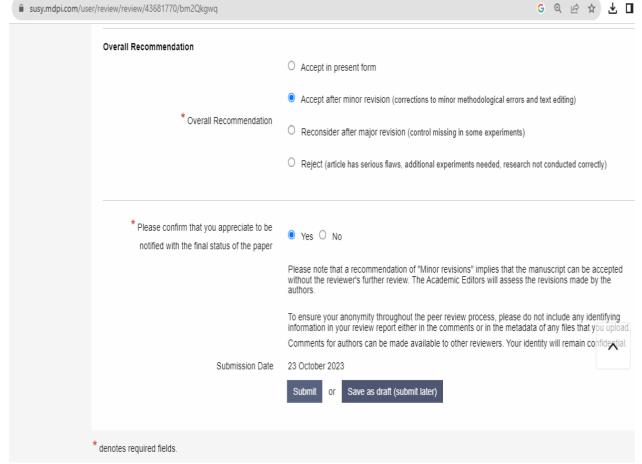
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Article Climatic Relationships to Agricultural Insurance Loss for the Pacific Northwest Region of the United States

Erich Seamon ¹^(b), Paul E. Gessler ¹^(b), John T. Abatzoglou ²^(b), Philip W. Mote ³^(b) and Stephen S. Lee ¹

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Abstract: Agricultural crop insurance is an important component for mitigating farm risk, particularly 1 given the potential for unexpected climatic events. Using a 2.8 million nationwide insurance claim 2 dataset from the United States Department of Agriculture (USDA), this research study examines 3 spatiotemporal variations of over 31,000 agricultural insurance loss claims across the 24-county region 4 of the inland Pacific Northwest (iPNW) portion of the United States, from 2001 to 2022. Wheat is the 5 dominant insurance loss crop for the region, accounting for over 2.8 billion dollars in indemnities, 6 with over 1.5 billion dollars resulting in claims due to drought. While fruit production generates 7 considerably lesser insurance losses (400 million dollars) as a primary result of freeze, frost and 8 hail, overall revenue ranks number one for the region, with over 2 billion dollars in sales. Principal 9 components analysis of crop insurance claims showed distinct spatial and temporal differentiation in 10 wheat and apples insurance losses using the range of damage causes as factor loadings. The first two 11 factor loadings for wheat account for approximately 50 percent of total variance for the region, with 12 apples having 60 percent variance. 13

Keywords: Pacific Northwest; agricandre; insurance; wheat; apples; drought

1. Introduction

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Crop insurance is an important component for mit The implications of the research have 1996 the United States Department of Agriculture (USDAnot presented explicitly. There must be Agency (RMA), which works to increase the availability a how the output of this study contributes surance as a risk management tool. With the implementa<mark>, to theoretical and practical progress in</mark> Act (FCIA) and the USDA RMA, program improvement the direction of future research farmers, implementing subsidies) grew the level of progradevelopment in to over 90 percent of all U.S. farmed land by 1998. Crop insurance program efforts have also had a dramatic impact on overall farm management, including the reduction of income risk around crop 23 production, increasing land values, increasing farm survivability rates, stabilizing cash 24 flow, and liquidity improvement [4]. By 2021, the USDA insured over 400 million acres of 25 farmland, with an insurance liability net worth almost 200 billion dollars [5]. Regionally, 26 agriculture in the Pacific Northwest (PNW) accounted for over 600,000 jobs over the three 27 state region of Idaho, Oregon, and Washington [6–8]. All three states consistently rank 28 in the top five in terms of U.S. crop production for a range of agricultural commodities, 29 including apples and wheat (Washington), potatoes and barley (Idaho), as well as hay, 30 blackberries, and hazelnuts (Oregon) [9]. In terms of agricultural exports, Washington 31 ranks second behind California (2021), with Oregon placing eighth and Idaho, eleventh [10]. 32 While indemnities and overall program costs have increased considerably since 2000, loss 33 ratios (a measure of total indemnities to total premiums) since the late 1990's have leveled 34 off at around 80 percent, mainly due to mandatory participation stipulations, underwriting 35 changes, and other legislative changes [5]. Given these combined efforts of 1) insurance 36 protection, as well as 2) risk mitigation (e.g. agricultural practices) which provide farmers 37

Citation: Seamon, E.; Gessler, P.E.; Abatzoglou, J.T., Mote, P.W., Lee, S.S. Climatic Relationships to Agricultural Insurance Loss for the Pacific Northwest Region of the United States. *Agriculture* **2023**, *1*, 0. https://doi.org/

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Copyright: © 2023 by the authors. Submitted to *Agriculture* for possible open access publication under the terms and conditions of the Creative Commons Attribution (CC BY) license (https://creativecommons.org/licenses/by/4.0/). protection against unforeseen natural disasters and economic events, our research focus is ³⁸ twofold: 1) to evaluate the variations of agricultural insurance loss for top commodities ³⁹ for the Pacific northwest (PNW) as well as the subregion of the inland Pacific Northwest ⁴⁰ (iPNW) (Figure 1), and 2) to examine how these variations align with climatically associated ⁴¹ causes of damage using dimensionality reduction and clustering methods. ⁴²

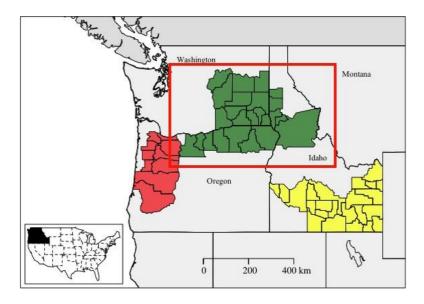


 Figure 1. Key agricultural regions in the Pacific Northwest (PNW) portion of the United States, with

 -county inland Pacific Northwest (iNPW) study area in ASUS

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Weather and climate extremes, including those at Include the source of Figure 1. Even though direct impacts on food security and resilience [11,12]. For is an elaboration or creation of the to a number of factors, including crop type, geograph authors, it needs to be entered in this Previous studies have examined climate-yield relatig way readers can know the clear source analyses examining climatic relationships related to cruform.which this image was compiled - --in particular, plays an important role in the success or failure of many agricultural systems. 48 Redmond [19] conceptually defines drought as "insufficient water to meet needs", with a 49 particular note of the varied relationships of supply and demand. Wilhite and Glantz [20] 50 describe drought broadly as a "deficiency of precipitation that results in water shortage for 51 some activity or for some group" and emphasize the difficulties in having one overarching 52 definition of drought, given its impacts from an agricultural, climatological, meteorological, 53 atmospheric, hydrologic, and water management perspectives. Operationally, drought is 54 often times quantified in terms of frequency, severity, intensity and duration, compared to a historical time frame, with human, biological, and climatological influences on both 56 water supply and demand. Typically referred to as a "creeping phenomenon", the impacts 57 of drought on society can persist for a number of years, dependent upon the level of 58 vulnerability [19]. Agriculturally, drought often refers to a period with anomalously low 59 soil moisture that substantially limits crop production [21]. Drought related impacts are 60 evident in agricultural insurance loss claims, both nationally as well within the PNW. For 61 example, drought conditions in 2015 resulted in agricultural insurance losses for PNW 62 wheat alone totaling 183 million dollars, with total financial losses for all commodities 63 ranging between 633 million and 773 million dollars [6]. 64

Grain-based cropping systems are particularly impacted by increased temperatures. Considerable research has examined the range of temperature impacts on grain yields [22] indicating that progressive temperature increases may initially result in increased yields, with an accelerating decrease over time, given an inverse temperature/precipitation re-

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lationship [23]. While increased temperatures will likely decrease wheat yields in the 70 region, the effects of carbon dioxide fertilization may modestly offset these yield reduc- 71 tions over time. In contrast, Schlenker and Roberts [14] suggest that yields for alternative 72 forms of cropping systems, such as soybeans, corn, and cotton, would slightly increase 73 with initial temperature increases up to 32 degrees Celsius, and then sharply decrease 74 as temperatures rise above that threshold. To make matters more complex, Rezaei et al. 75 [24] as well as Asseng et al. [25] indicate that unique cultivars within a species may have 76 varying phenological cycles, suggesting that any agricultural climate impacts assessment 77 should include a variety of sub-species for proper threshold analysis. When examined in total, climatic relationships to agriculture are extremely variable, with changing out- 79 comes due to cropping system, regionalization, farming practices, and genetic diversity. 80 This complexity is encapsulated in agricultural insurance loss management, in order to 81 effectively hedge agricultural risk, associated variability and complexity, and incorporated into a time-adjusted financial premium/payout process. Under this premise, evaluating 83 insurance losses in relationship to sub-seasonal climatic impacts provides a reasonable 84 approach to assess patterns and predictability, without delving into the underlying crop 85 processes and their biophysical effects due to a changing climate.

From a seasonal perspective, adverse growing conditions (such as during drought) can force 88 farmers to consider additional risk management approaches that complement insurance 89 mechanisms, including irrigation, selective crop abandonment, crop diversification, as 90 well as unique crop rotation practices, which may mitigate current and future losses and 91 preserve long-term economic viability of cropping systems [26,27]. For example, crop 92 producers who utilize conservation tillage are often able to improve the capture and 93 storage of soil moisture, which provides their crops an important buffer against drought 94 impacts. By increasing the number of crop types as part of a rotation cycle, altering seeding 95 dates, as well as using drought-sensitive breeds, farmers can retain more available soil 96 moisture (reducing long term drawdown), while maximizing production and sales by 97 spreading risk across a larger set of commodities [28]. From an adaptive perspective, the 98 economic implications of more severe drought conditions, as well as changes in drought 99 characteristics, may encourage farmers to consider alternative crop systems that are more 100 economically viable. In total, these added risk management efforts, in combination with 101 crop insurance, provide farmers with a diversified ability to mitigate potential financial 102 loss in the face of changing economic and climatic conditions. 103

Given the spatial diversity in terms of cropping systems across Idaho, Oregon and Washing-105 ton, the iPNW sub-region provides a more homogeneous, well distributed dryland farming 106 region, allowing us to explore spatial and temporal variations, while maintaining a fairly 107 consistent county level claim total across the area as a whole. This narrowing also allows for 108 the elimination of counties where little or no insurance claims were filed, primarily due to 109 landscape, urbanization, or profitability constraints. From a damage cause perspective, the 110 focus is on losses due to weather and climate extremes, particularly those due to drought 111 and heat (wheat) and freeze, frost and hail (apples). _112

2. Materials and Methods

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The USDA's data archive of agricultural insurance 1989 to 2022, was the primary dataset for this analysis ance claims provided at monthly temporal and county la record represented a unique claim associated with a fa amount of the insured loss, the commodity type relat canola), the acreage for the loss, the insurance compa most notably, a cause for the crop damage (e.g. heat, dra	The research gap has not been clearly ASUS 2023-11-02 12:13:16 As far as the reviewer knows, it is no past unethical binclude citations from Uterly

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of irrigation supply). The extent of this data archive is considerable: for example, from 121 1989 to 2022, the USDA's crop insurance data collection for the United States (all commodi-122 ties) totals approximately 2.8 million claims, with 31,000 claims originating in the Pacific 123 Northwest (Idaho, Oregon, and Washington) for over 35 different commodities, across 30 different damage causes. 125

For our analysis, we construct a basic three step analysis methodology which allows us to examine commodity-specific insurance loss across damage causes. Given our research goal to examine iPNW spatiotemporal variation of agricultural insurance loss, the results of these steps not only permits us to narrow our factorial analyses by geography, time, commodity, and damage cause, but also enable comparisons of how water scarcity (drought and heat) and water excess/cold (freeze, frost, hail) damage causes vary based on commodity type and geography.

We initially perform a full examination of insurance loss across all commodities and 135 damage causes, for the entire PNW region, from 1989 to 2022. As part of this step, we 136 aggregate the data by county, commodity, year, and damage cause. An initial data review 137 indicates that approximately 83 percent of insurance loss for the region occurred after 2000 138 (Supplemental Figure S2), which comports with farm bill policy incentives implemented 139 in 1998, increasing crop insurance participation (acres) to over 90 percent [5]. Across the 140 three state PNW region, over 75 percent of insurance losses occurred within the iPNW, 141 with wheat losses being the overwhelming dominant commodity. In addition, acreage data 142 was not recorded for individual claims until after 2000 as well. As such, we limit our time 143 frame of insurance loss examination to 2001 to 2022 and narrow our study area region to 144 the 24-county region of the iPNW (Figure 1). This reduction of data by year additionally 145 helps to resolve missing data issues in some counties that have no insurance claims, and 146 thus no revenue loss. 147

We then use principal component analysis (PCA) to identify commonalities in insurance claims across years, counties, commodities, and damage claims in the iPNW. PCA is a data dimensionality reduction technique which computes a new set of variables by maximizing the variance of all input variables, and then examines the linear combinations of said variables in orthogonal space [29,30]. PCA notation can be described as follows: 149

$$\boldsymbol{\alpha}_{k}^{\prime} \boldsymbol{x} = \sum_{i=1}^{p} \boldsymbol{\alpha}_{k}^{\prime} \boldsymbol{j}^{x} \boldsymbol{j} \tag{1}$$

Where: *x* is a vector of random variables (*p*) $\boldsymbol{\alpha}_k$ is a vector of p constants

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The process is to initially find a linear function of $(x, \alpha'_1 k)$ with a maximum variance. Next, we find another linear function of $(x, \alpha'_2 k)$ which is uncorrelated with the maximum variance of $(x, \alpha'_1 k)$. The approach is iterated over the extent of available variables. Ideally the most variation in x will be accounted for by *m* principal components where m < p. 157

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Given the nested structure of the data (insurance claims by county, year, commodity, and damage cause), we construct a multitude of principal components analyses (with damage cause insurance loss totals (U.S. dollar) as our factor loadings), using differing combinations of county, commodity, month, and year (county by year, county by month, and county by commodity), for both the entire PNW three state area, as well as for the wheat growing 166

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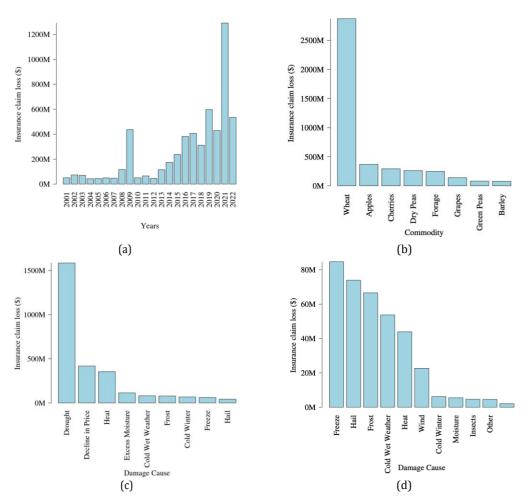
region of the iPNW (Supplemental Figures S15 - S22). The full range of PCA outputs are 167 provided in our supplemental materials. Using this approach, we create a set of input 168 variables for our PCA, to examine how damage cause factors were associated, as well as 169 how counties and years were aligned to these individual factor loading vectors. In order to 170 evaluate how PCA variables group together, we apply a kmeans algorithm method [31] to $_{171}$ estimate optimal clusters (based on Euclidean distance) for both county and year, based on 172 our PCA outputs. Kmeans clustering is a vector quantization method which maps input 173 values from larger to smaller sets. By iteratively partitioning *n* observations into a known 174 set of clusters, the kmeans algorithm attempts to converge on an optimum grouping of 175 clusters, based on a common spatial extent. This two-step clustering analysis has been 176 noted as an effective approach in combining dimensionality reduction with unsupervised 177 learning methods [32]. 178

From the results of our initial data inspection and kmeans-applied PCA, we limit our 180 commodity analyses to wheat and apples, and narrow our set of damage cause claims to 181 areas of water scarcity (drought and heat) as well as water excess/cold (freeze, frost, and 182 hail). We then examine losses for the region, exploring temporal and spatial relationships 183 on an annual basis. In addition, we compare insurance loss with overall commodity 184 production across the 24-county study area from 2001 to 2022. 185

3. Results

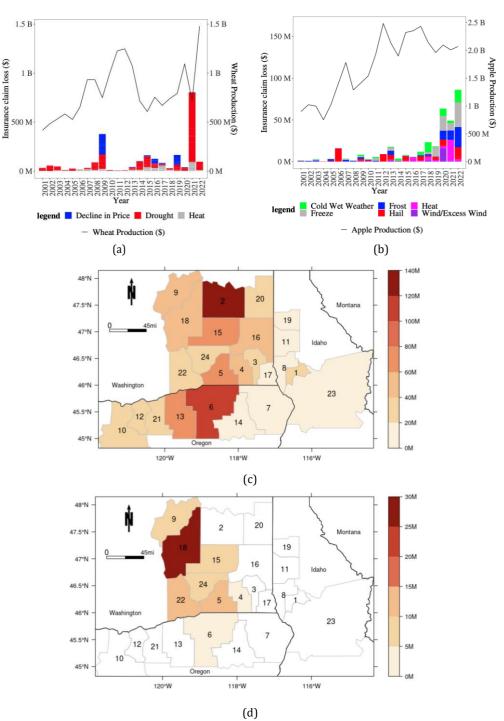
PNW insurance claims from 2001 to 2022 totaled over 33,000, for all commodities, 187 with overall insured losses of ~ 6.5 billion dollars. Wheat, the dominant commodity for 188 insurance claims in the three-state region, accounted for approximately 20,600 filings, with 189 total losses of 3.5 billion dollars for the same time period. Apples and cherries were a 190 distant second and third in terms of overall losses (Supplemental Figure S5), each with 191 approximately 600 million dollars, with potatoes and peas adding a minimal contribution 192 to the overall total (250 million dollars each). Narrowing our analysis to the iPNW, we 193 see that insurance losses there made up approximately 72 percent of the total amount of 194 loss for PNW as a whole. Wheat was similarly the predominant commodity incurring 195 insurance loss for the iPNW, with over 2.5 billion dollars in claims, with apples coming 196 in a distant second, at 325 million dollars. In term of damage cause, drought resulted in 197 the largest amount of insurance loss for the PNW overall, at over 1.8 billion dollars, with 198 decline in price (850 million dollars) and heat (800 million dollars) coming in second and 199 third, respectively. Focusing in on the iPNW, the leading damage causes for this region 200 were drought and heat, which combined to account for approximately 2.65 billon dollars in 201 losses from 2001 to 2022. For all commodities, drought and heat-related claims for the iPNW 202 accounted for 68 percent of all insurance losses in total for the 2001 to 2022 time period. 203 There was additionally considerable variability across iPNW crop types with regards to 204 damage-specific insurance claims. For example, wheat insurance losses were dominated 205 by drought and heat, with apples and cherries claims aligned with freeze, frost, and cold 206 weather (Figure 2). 207

In order to address our research questions around spatial and temporal variations of 209 insurance loss related to water availability, we narrowed our commodity analysis to apples 210 and wheat, the two dominant commodities for the region. Annual wheat losses specifically 211 due to drought, heat, and excessive moisture for the iPNW were analyzed for each year in 212 the period from 2001 to 2022, while apples were examined for the same period, focusing on 213 freeze, frost, and hail. Our results for this 2001-2022 time period show that the year-to-year 214 variation of losses for wheat are dominated by drought, with peak years of 2009 and 2021. In 215 contrast, 2011 had almost no drought or heat insurance losses, with excessive moisture and 216 rain being the dominant damage cause factors. This annual variability aligns with historical 217



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climatological variations. While 2011 was a particular wet year for the PNW [33], 2021 218 experienced a significant drought primarily attributed to extreme summer temperatures 219 during a two week window in June and early July. This event resulted in the highest 220 recorded mean summer near-surface air temperatures for the PNW from 1950 to 2021 [34], 221 which is evident in the more than double annual wheat insurance losses, in the range of 700 ²²² million dollars (Figure 3). When decline in price is incorporated into this annual view for 223 wheat, we see certain years where a large majority of claims are associated with economic 224 decline; for example, in 2009, decline in price claims align with wheat prices declines from 225 430 dollars /metric ton to 220 dollars /metric ton. Wheat production varies inversely with 226 losses, with the lowest levels of production occurring in years with the highest levels of 227 drought/heat insurance loss. Comparatively, apple insurance loss for the region shows a 228 more gradual increase from 2001 to 2022, with 2020-2022 having a considerable increase 229 in freeze/frost/hail losses. Apples show a peak loss year of 2022, which coincides with 230 relatively lower losses for wheat associated with drought and heat, during the same time 231 frame. Additionally, apple losses, while not typically effected by heat/drought events, still 232 had relatively large losses in 2021, which is a testament to the severity of 2021 drought/heat 233 impacts across many commodities. Unlike wheat, apple production is roughly 15 times 234 larger than insurance loss claims, which may have associations with economic systems, 235 as well as water availability influences (e.g. drought may have a much greater impact on 236 insurance claim submittals vs. freeze/frost/hail claims). 237



1 Lewis: 2 Lincoln: 3 Garfield: 4 Columbia: 5 Walla Walla: 6 Umatilla: 7 Wallowa: 8 Nez Perce: 9 Douglas: 10 Wasco: 11 Latah: 12 Sherman: 13 Morrow: 14 Union: 15 Adams: 16 Whitman: 17 Asotin: 18 Grant: 19 Benewah: 20 Spokane: 21 Gilliam: 22 Benton: 23 Idaho: 24 Franklin

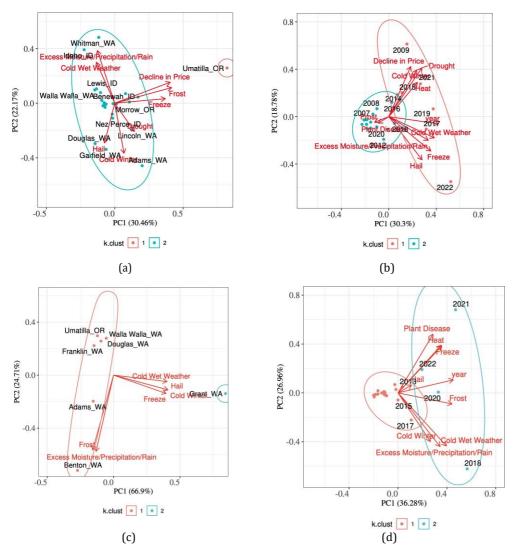
Figure 3. a) Stacked barplot of losses from 2001-1022 for wheat: b) stacked_barplot of_losses_from_____ 2001 22 for apples; c) map of wheat insurance loss due to h insurance loss due to cold weather, freeze, frost, hail, heat, and 2023-11-02 12:14:23 The source of the analysis data must be

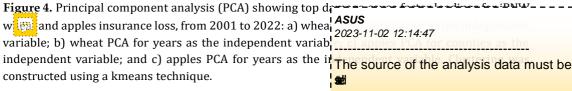
Spatially, while total 2001 to 2022 wheat losses (manage causes) were highest in 238 Adams county, WA (232 million dollars), wheat insurance loss due to drought and heat 239 were highest in Lincoln county, WA (133 million dollars), as well as counties along the 240 northeastern portion of the Oregon high desert (Umatilla county, OR at 119 million dollars 241

and Morrow county, Oregon at 68 million dollars). From a percentage breakdown, over 50 242 percent of all damage cause losses in Umatilla were a result of drought/heat, with over 243 40 percent attributable to drought/heat in Adams and Lincoln counties, Washington. If 244 we specifically examine spatial differences in wheat drought/heat insurance loss by year, 245 we see notably different patterns of loss concentrations between 2009 and 2021. For 2009, 246 the region's few drought and heat claims were concentrated in the north central portion 247 of the region, with losses in the highly productive Columbia river region being relatively 248 low. In contrast, 2021 wheat losses due to drought and heat were concentrated in the 249 upper portion of the Washington Palouse region (Whitman, Lincoln, Adams, and Douglas 250 counties), with additional loss concentrations falling along the Columbia river valley and 251 in the western portion of the Palouse (Figure 3). In order to better understand the factorial 252 relationships of damage causes, two principal component analyses were run for the iPNW 253 region for both wheat and apples, to explore 1) spatial (county) as well as 2) temporal (year) 254 variation. Both PC analyses use damage causes as the factor loadings, with all data scaled 255 by the unit variance. Additionally we use singular value decomposition (SVD), a form of 256 matrix factorization which is considered a superior method for PCA computation [35]. For 257 wheat by county, approximately 53 percent of total variance of insurance loss by county 258 level damage cause can be attributed to the first two principal components, with water 259 scarcity (drought/heat/fire) damage causes having a negative coordinate alignment in 260 terms of the first principal component (PC1) vector loading directions. For apples by county, 261 over 90 percent of total variance can be attributed to the first two principal components, 262 with excessive water and cold-related damage causes. When we examine variation by 263 year, we see less explained variance for the first two principal components, with wheat 264 accounting for 48 percent explainability, and 62 percent for apples (Figure 4). Examining 265 PC loadings by county, we see a clear alignment of water scarcity damage causes in highly 266 productive wheat counties (Umatilla county, OR, Lincoln and Whitman counties, WA), with 267 orthogonal damage causes (excessive moisture/freeze/frost) aligning with counties that 268 are typically in highly productive fruit production regions (e.g. Grant and Benton counties, 269 WA). Applying a kmeans clustering algorithm with an elbow cluster optimization selection 270 method, we identified two key clusters in the two-dimensional PCA space, that additionally 271 support the differentiation of water scarcity PC1 loadings from PC2 water excess. When 272 PCA was run using year as the independent factor (2001 to 2022) and applying a kmeans 273 clustering algorithm with an elbow cluster optimization selection method, we identified 274 two key clusters. The identified clusters support the differentiation of water scarcity PC1 275 loadings from PC2 water excess. Most notably, 2009, 2018, and 2021 are within a distinct 276 cluster falling along damage cause groupings for drought, fire, and heat (Figure 4). 277

4. Discussion

Given ou every data analysis to examine ASUS 2023-11-02 12:17:54 of insurance loss in relationship to climatic damage ca unique spatial and temporal patterns that appear to al Discussions must be described trends. The considerable crop-specific variations in ter considerable crop-specific variations in ter effects due to drought and heat, vs excessive freeze/cexisting findings with past research. a clear and straightforward signal for generalized clin<mark>tThat way, it produces a significant gap</mark> with crop insurance fluctuations. As previously noted, the difference The relevance and and heat claims for 2021 closely align with the extreme summer heat event in the PNW, 286 which effect not only cereal systems, but also impact fruit commodities. In addition, such 287 patterns provide an important perspective on climate variability vs. economics, and the 288 sensitivities of agricultural systems to differing effects. Of particular interest were the 289 differences in iPNW wheat insurance loss, comparing 2009, 2021, and 2022, in terms of 290 the drought, heat, excessive moisture, and decline in price total losses. While 2009 and 291 2021 have large dollar losses in terms of drought and heat, 2022 additionally had relatively 292 larger values with regards to cold weather, rain, and freeze. While increased drought and 293 heat losses in 2021 align well with regional drought conditions [36], increased drought 294





and heat claims for 2009 seem to conflict with comparable climate conditions for that year, 295 which indicates that the iPNW was not in a period of drought-[37]. These insurance los comparisons between 2009 and 2021 suggest that, in compromised economic conditions (e.g. price decline), claims due to climatic damage causes may increase, even though 298 actual climatic conditions do not warrant such increases [38,39]. This may also indicate that particular commodity-specific thresholds exist where economic factors dominate over 300 climatic impacts, resulting in a broad distribution of claim loss across a range of damage 301 causes. 2011 losses were interestingly juxtaposed to 2009 and 2021, with very little drought 302 or heat insurance claims, but with the largest amount of excessive moisture filings of any 303 year in the period of analysis. Additionally, we saw an inverse relationship between annual 304 wheat production and drought/heat insurance loss, with 2021 being the only year in this 305 time period where losses were higher than production. Work by Quiggin et al. [40], Miranda 306 and Glauber [3], and Glauber [41] all reference the relationships of insurance loss with 307 overall crop production, supporting this inverse relationship scenario. Spatial variations 308 of wheat insurance losses due to drought and heat provide an additional perspective in 309

terms of locational sensitivities to climate. With variations of phenology, claim frequency, ³¹⁰ regional crop development, irrigation, and cropping practices, commodity-based insurance ³¹¹ claim analysis for agriculturally homogeneous regions may provide the best framework for ³¹² delineating differences in claim/loss variation, based on time and the cause of damage. ³¹³

5. Conclusions

ASUS The distinct differences in annual variation, as w 2023-11-02 12:20:07 suggest insurance loss analysis may serve as a more effective of influences [42]. Our results additionally highlight that I suggest creating 1 new paragraph aspects of climate and economic impact together (e.g. origination of policy damage causes), given that farmer decisions regarding recommendations and the weaknesses typically take into account these two factors simultaneo, of this research. Limitations of the study to file a crop insurance claim depend upon a multitude must be revealed to provide broader which may be directly or indirectly impacted by extreminsight into future research ogical events [43]. For example, during economically stable perimprovements commodity prices), a farmer may be disincentivized to file a drought-associated claim, particularly given the balance between production value and insurance payout. Conversely, during periods of economic instability when commodity prices may be declining, farmers may be incentivized 326 to initiate a claim in periods of moderate drought. 327

The results of this work highlight several areas of potential future research, par-329 ticularly around understanding the interactions between insurance loss, conservation 330 practices, economic factors, climate influences, and policy effects, as well as regional dif-331 ferences/similarities of damage cause influences across a range of commodities other 332 than wheat. Under changing climate and conservation practice conditions, there may 333 be situations where crop insurance risk management may incentivize, or disincentivize, 334 farm practices that reduce agricultural climate change impacts, given their individualized 335 economic implications. Additionally, this work may assist future research in identifying 336 the financial impacts of a changing climate on insurance loss, over time and differing 337 geographies. 338

Supplementary Materials: All supporting information and reproducible code for all figures can be downloaded at: https://doi.org/10.5061/dryad.hhmgqnknh 340

Author Contributions: For research articles with several authors, a short paragraph specifying their341individual contributions must be provided. The following statements should be used "Concep-342tualization, E.S, and P.G.; methodology, E.S.; software, E.S.; validation, E.S.; formal analysis, E.S.343and P.G.; investigation, E.S.; resources, P.G. and P.M.; data curation, E.S.; writing—original draft344preparation, E.S. and P.G.; writing—review and editing, E.S., P.G., J.A., P.M., and S.L.; visualization,345E.S.; supervision, P.G., J.A., P.M., and S.L.; project administration, E.S.; funding acquisition, P.G. and346P.M. All authors have read and agreed to the published version of the manuscript.347

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3.	Miranda, M.J.; Glauber, J.W. Systemic Risk, Reinsurance, and the Failure of Crop Insu <i>Agricultural Economics</i> 1997 , <i>79</i> , 206–215. https://doi.org/10.2307/1243954.	it snecessary to add several citations related to the topics discussed.

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