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# The impact of production and price risk on ecosystem goods and services provision from agriculture and forestry in mountainous regions

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**Abstract:** Assessing the impact of climate change on Ecosystem Services (ES) requires an understanding of the dynamic link between land cover, ES and economically driven land use decisions. Land use decisions taken by farmers and foresters depend on the yield levels as well as on the prices for outputs and inputs the decision maker expect. However, the variability of yields and prices, i.e. income risk, is also highly relevant for the decision maker. Increasing volatility of climate and more volatile markets for commodities will enhance the relevance of risk for land use decisions in the future.

To assess the impact of increasing production and price risks, we develop an agent-based economic land allocation model. Applying discrete stochastic programming and a safety first approach we integrate the consideration of income variability in the agents' decision making.

We intend to apply our model to the temperature sensitive alpine region of Davos, Switzerland and to compare land use under the consideration of production and price risks as well as using a deterministic (i.e. non-risky) baseline.

**Keywords:** Land use change, Income variability, Agent-based modelling, Production risk

## 1 INTRODUCTION

In Swiss mountains, very few regions are still truly natural. Land use and the maintenance of Ecosystem Services (ES) provision is determined through a complex interplay between human action and basic ecosystem processes (Lambin et al. 2001). This is particularly the case if ES are provided in ecosystems that are managed to a large extent by humans. In mountainous regions, farmers and foresters are the two main management agents that influence ecosystems. They manage the land to pursue their own goals, such as achieving a sufficient income. Their decisions, however, are not only determined by expectations about average

54 incomes, but also by the impact of their actions on the income variability (e.g.  
55 Hardaker et al. 1997).  
56 Income variability in farming can have different causes (e.g. Moschini and  
57 Hennessy 2001) such as the uncertainty about product and input prices, or the  
58 uncertainty regarding the level of production that will be achieved in each year.  
59 Furthermore, crop yields can be influenced by variable weather both positively and  
60 negatively (Moschini and Hennessy 2001).  
61 In order to stabilize income, i.e. to mitigate risks, farmers may change their  
62 production using different risk-management strategies (Reidsma et al. 2010), for  
63 example the storage of grass to smooth the availability of fodder over multiple  
64 years (Mosnier et al. 2011). Furthermore, production intensities in fodder and crop  
65 production as well as the portfolio of agricultural activities used may change in  
66 presence of risks (Finger et al. 2010). Besides direct impacts on land use, risk-  
67 management strategies can also indirectly impact land use decisions. For example,  
68 the demand for feedstuff will be lower if the farmer mitigates risk by housing fewer  
69 animals (Ritten et al. 2010). Since the farmer still will cultivate the available  
70 farmland, he is expected to reduce intensity of cultivation or to introduce alternative  
71 activities such as crop production. Another risk management option that directly  
72 impacts land use is the expansion of forest areas. Since perennial cultures such as  
73 energy crops like switchgrass or even forests are less affected by annual volatilities  
74 this can reduce income variability (e.g. Bocqueho and Jacquet 2010).  
75 Based on this background, we aim to investigate the role of farmers' income risks  
76 on land use and ES provision under current and future climate and market  
77 conditions. Our analysis is based on the community of Davos located in the canton  
78 of Grison within the Swiss Alps. This example has been chosen because most  
79 mountainous ecosystems in Switzerland are managed by farmers and foresters,  
80 i.e. the impact of their decisions on ES provision in these regions is significant. In  
81 addition the assessment of changes in land use due to increased variability is  
82 especially important since these changes may determine ES provision, which are  
83 highly important in these regions.  
84 The assessment of driving forces for land use change in mountainous regions,  
85 such as the community of Davos, must rely on highly complex and integrative  
86 methods since these regions are characterized by a high degree of heterogeneity  
87 and complex topography (Briner et al. 2012). For instance, steep slopes that  
88 require specialized machinery and manual labor, both significantly increasing the  
89 costs of production. In mountain agriculture costs are further increased and  
90 potential revenues (e.g. from yields) are further decreased due to poor soils,  
91 heterogeneous land suitability and the presence of a harsh climate (Tasser et al.  
92 2011). If land use change is simulated in mountainous regions it is therefore  
93 mandatory to consider the high spatial variability.  
94 Current studies on local and regional land use change consider only the impact of  
95 climate change and socio-economic changes on mean values. As such they  
96 consider the impact of climate change on changes in the average yields of grass-  
97 and cropland and not on changes in variability (see for example Henseler et al.  
98 2009, Briner et al. 2012). These models neglect that different decision makers have  
99 an unequal ability (and preference) to bear risk (Irwin and Geoghegan 2001). In  
100 contrast, farm models that overtly consider variability are not spatially explicit (e.g.  
101 Mosnier et al. 2011). We address this deficiency by expanding the agent-based  
102 land-allocation model ALUAM-AB (Huber et al. 2012) in a way that the impact of  
103 income variability in the goal function of the different agents can be considered. We  
104 then apply this model to the region of Davos to answer the following research  
105 questions:  
106 • What is the impact of income variability (due to stochasticity in production or  
107 prices) on land use?  
108 • How will land use change if agents show different degrees of risk aversion?  
109 To answer these questions this article proceeds as following. First the agent based  
110 model is described with an especial focus on the risk assessment part. Then  
111 expected results are described. In the end possible impact of land use change on  
112 ES provision will be discussed.

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## 2 METHODS

For the assessment of the impact of uncertainty in weather and prices we expand earlier work (Huber et al. 2012). More specifically, we use a recursive dynamic agent based model to increase the understanding of agricultural land use changes triggered by market and policy changes considering individual preferences of the farmers.

### 2.1 Land use activities

The land use activities considered follow Briner et al. (2012). We consider 9 different grassland activities, 3 cropping activities and one forest activity. Grassland activities differ in the way the grass is harvested (grazed or mown), by the decision if grassland activities are integrated in a crop rotation and in the intensity of cultivation of the grassland. Available cropping activities are wheat, barley and potato, each at one intensity level. In addition agents have the possibility to use the land as forests. Weather does not have an immediate impact on the yield of forests, since forest yields are climate but not weather dependent. This means that yearly favourable and unfavourable weather is compensated during the long growing period. In contrast to agricultural land uses forest is therefore assumed to be a risk free alternative with constant yields. Note that this assumption is also based on the employed discounting. Thus, the occurrence of risks (losses) in forest production in 50 years may not be as relevant for the decision maker as those in agricultural production occurring today, while agricultural production, however, is assumed to differ between years.

### 2.2 Livestock activities

The farmers represented in the model have the possibility to house different ruminants limited to the production systems currently relevant in the region. Available are dairy cows, suckler cows, heifers, sheep for milk production, sheep for meat production. One housing system is available per animal species except for dairy cows, which can either be housed in a stanchion barn or in a loose barn. To minimize aggregation bias, the housing systems were modeled in different sizes allowing different levels of labor productivity and investment costs per animal.

### 2.3 Balances

#### 2.3.1 Fodder balance

The demanded fodder can either be produced on farm or – in certain scenarios – can be purchased on a roughage market at common prices. If additional fodder is produced, but not needed for feeding livestock, it can be sold on the market. Prices for purchased fodder are much higher than for sold fodder. For each farm (i.e. each agent) and year, an equal fodder balance is required. Note that concentrate feedstuff cannot be produced in the region due to unfavorable climatic conditions and therefore needs to be bought by the farmer.

#### 2.3.2 Nutrient balance

To receive direct payments farmers have to fulfill certain cross-compliance restrictions (Proof of Ecological Performance PEP, see El Benni and Lehmann 2010 for details). One of these restrictions is limiting the available nutrients from livestock activities and artificial fertilizer. This nutrient balance is calculated based on the official Swiss methodology (Suisse Bilanz; see Amaudruz et al. 2003 for

172 details). The amount of available plant nutrients in the model therefore is not  
173 allowed to exceed demand by land use activities by more than 10%. To maintain  
174 adequate yields and soil fertility minimum available plant nutrients must meet  
175 demand calculated for each land use activity on each parcel. If there are not  
176 enough nutrients available from livestock activities agents have the possibility to  
177 purchase additional artificial fertilizer.

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### 2.3.3 Labor balance

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181 The available amount of labor needs to be enough to accomplish all farm-work,  
182 including farm management. To ensure a sufficient amount of labor available  
183 during the whole year, two different balances are calculated. First labor balance is  
184 spanning over the whole simulation period of one year. Since in mountain regions  
185 the time that can be used for conservation of fodder for the winter season is only  
186 very short this results in a distinct peak in work load during summer season. To  
187 ensure that there is enough labor available during the vegetation period there is a  
188 second labor balance calculated for this specific period.

189 The available labor is mainly the farmer and his family, for whom different  
190 opportunity costs for labor are assumed. In addition to family labor, there is also a  
191 possibility to hire additional labor on an hourly rate.

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## 2.4 Calculation of the land-rent

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196 Calculation of the current costs for inputs is based on price statistics as provided by  
197 Blum et al. (2010). Since we focus on output price uncertainty, inputs' prices are  
198 expected to remain at a constant level. In contrast, yearly prices for outputs are  
199 drawn randomly from a normal distribution. This distribution is estimated using past  
200 price data.

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## 2.5 Design of the agents in the model

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205 The design of the agents is based on information from agricultural census data. All  
206 running farms in the region are divided into 10 different clusters based on their  
207 production structure, which is formed by the production system and number of  
208 livestock housed on the farm, the size of the land cultivated and land use. All farms  
209 falling inside one of these 10 clusters are then aggregated to one agent. For the  
210 calculation of the available workforce the number of farmers is also aggregated.  
211 However we explicitly account for the age of all farmers in the model, to calculate  
212 the number of farmers that retire after each decision period.

213 Further agent characteristics are based on a survey made with farmers in another  
214 mountainous region of Switzerland, on literature (Rossier and Wyss 2006) and on  
215 expert knowledge. Agents were characterized with respect to their available time to  
216 work on the farm, for their expected minimum income they need to achieve to keep  
217 the farm running and for their risk preferences.

218 Since there are no spatially explicit data about the land ownership, available  
219 parcels are randomly distributed among agents. The implementation of several  
220 restrictions guarantees that the share of land of each quality, represented by the  
221 slope, is consistent with reality.

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## 2.6 Optimization procedure

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226 The particular focus of our contribution is that the decision making of the farmers  
227 takes into account expected income as well as income variability. We thus apply a  
228 combined discrete stochastic programming (DSP) and safety first-approach (e.g.  
229 Cocks 1968). The strength of DSP is that it makes possible solving multi-stage  
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231 decision problems with uncertain outcomes. It is possible to introduce uncertainty  
 232 into the coefficients of the objective function as well as into input-output coefficients  
 233 or resource endowments. It is only necessary to know the probability of realization  
 234 of different discrete states of nature, which makes it possible to directly use yield  
 235 data generated with a crop-yield model as used in this modeling framework.  
 236 We use a two stage decision model. In the first stage, the land use activities are  
 237 chosen for all different years. Based on the results of this decision in the second  
 238 stage livestock activities are optimized, i.e. the number of different animals is  
 239 adapted in order to reach maximum income. In addition, in stage two decisions  
 240 about trade of fodder as well as the hire of additional labor are made. In this  
 241 framework, it is assumed that there is complete information about the outcome of  
 242 the first stage decision (Appland and Hauer 1993). This means that the farmer  
 243 already knows the outcome of this land use decision, i.e. the amount of grass  
 244 harvested, when he decides how many animals he wants to house etc.  
 245 The optimization procedure is described in equations 1-4. Equation 1 shows the  
 246 goal function that is optimized to maximize aggregated land rent  $Z$ .  $X$  represents  
 247 the decision vector of the decision of stage 1, i.e. the decision of the choice of the  
 248 land use activities. These decisions are made on an annual basis. The decision  
 249 vector  $Y$  is the same for all different states of nature  $k$ . These states of nature  $k$   
 250 differ in weather and subsequently in the yields that can be harvested on crop- and  
 251 grassland.  $g$  represents the profit margin (i.e. win or loss) attributed to each land  
 252 use activity.  $Y$  is the decision vector of decisions made in stage 2. This is the  
 253 choice of the livestock system, the number of animals housed as well as decisions  
 254 about the trade of fodder. Since farmers are assumed to know the outcome of their  
 255 land use decisions, they can adapt to changes in yield and prices. Decision vector  
 256  $Y$  is therefore dependent on the state of nature  $k$ .

$$\max Z_t = \frac{1}{k} \sum_{k=1}^{10} gX_t + m_{k,t}Y_{k,t} \quad (1)$$

$$\text{s.t.} \quad aX_t + bY_{k,t} < r \quad (2)$$

$$c_kX_t - dY_{k,t} > 0 \quad (3)$$

$$X_t \geq 0, Y_{k,t} \geq 0 \quad (4)$$

257 Optimization is subject to different restrictions described in section 2.3. These are  
 258 represented by Equations 2 to 4. Constants  $a$  and  $b$  represent demand for  
 259 resources of the different activities.  $r$  is the amount of resources available.  $c$  and  $d$   
 260 are supply and demand of fodder or nutrients that are transferred between land  
 261 use activities and livestock activities. For details about the applied restrictions see  
 262 section 2.3.

263 In a subsequent step a time dimension  $t$  is introduced into the model, i.e. the model  
 264 is made recursive dynamic. Temporal dynamics are driven through changes in  
 265 economic parameters that differ between years. Recursive dynamic elements are  
 266 the number of stable houses, the number of farmers represented by one agent in  
 267 the model, and the land distribution.

268 Since we assume that farmers do not only optimize expected income but also take  
 269 into account income variability in a next step a safety-first approach is implemented  
 270 in the modelling approach. This approach is based on Roy (1952) and since then  
 271 was used in several studies (e.g. Haley 2012). In its original version the safety-first  
 272 approach means that farmers intend to optimize their product portfolio in a way that  
 273 the probability that income is lower than a certain threshold is minimized. In this  
 274 study we apply a more restrictive version of the safety first approach in that we  
 275 assume that a certain threshold depending on the agent's risk aversion is not  
 276 allowed to be undercut since the farm goes bankruptcy otherwise.

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## 280 2.7 Land-market

281 Different agents are linked over a market for land as suggested by Lauber (2006).  
 282 Land enters the market only if it is not cultivated by the agent anymore. This can be  
 283 the case because cultivation is not profitable anymore, i.e. an agent lets a parcel  
 284

285 fall fallow. Furthermore, the retirement of a farmer without successor and giving up  
286 of a farm because income falls beyond a certain threshold can induce land to enter  
287 the market. Since the agents in the model represent an aggregation of several  
288 farms with a similar production structure they can retire partly each year. If a real  
289 farmer, who is part of an agent, becomes retired only the share of the agent's land  
290 corresponding to this farmer enters the market if there is no successor. All the  
291 other farms that are part of the agent stay with the agent. The probability that the  
292 single farmer has a successor is thereby dependent on the income that is possible  
293 to achieve on the farm as well as on the size of the farm (Rossier and Wyss 2006).  
294 The distribution of the land that enters the market is divided into two steps: 1) It is  
295 assessed which agents are interested in the land. This is assumed to be the case if  
296 the shadow price for the use of a certain parcel is positive, i.e. if the additional  
297 parcel would increase the income of the agent. 2) The available parcels are  
298 assigned randomly to the agents that are interested in these parcels.

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## 301 **2.8 Crop-yield model**

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303 Sets of ten years of projected future yields of relevant crops are calculated using  
304 data on optimal and absolute crop growing conditions (FAO 2007) for every parcel  
305 of 100m by 100m. The minimum and maximum temperature and precipitation  
306 values and the values that define the crops temperature and precipitation  
307 extremes, are extracted from the FAO crop data base EcoCrop (FAO 2007).  
308 Applying these values, we fit a relative crop yield curve for temperature and  
309 precipitation values using an incomplete beta distribution (Briner et al. 2012).  
310 These species specific crop yield curves are then used to calculate the relative  
311 yield for six crops based on monthly precipitation and temperature values. The  
312 projected realized yield is taken as the minimum yield value from the temperature  
313 and precipitation responses. The absolute yield of crops is calculated by  
314 standardizing the values against observed yield of crops in 2000.

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## 317 **2.9 Forest landscape model**

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319 The spatially explicit forest-landscape model TreeMig (Lischke et al. 2006) allows a  
320 grid-based simulation of forest dynamics over several centuries with different  
321 climate change scenarios. Each 100 m by 100 m cell simulates forest dynamics  
322 based on species specific germination, establishment, growth, competition,  
323 reproduction and mortality. Spatial interaction between the cells is given by explicit  
324 seed dispersal simulation. As an input to the land-allocation model, TreeMig  
325 provides information on potential changes in tree species diversity and forest  
326 yields, driven by climate change.

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## 329 **3 EXPECTED RESULTS**

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331 Based on the results of other studies that accounted for uncertainty in farmers'  
332 decision making, we expect that land use will be less intensive if income risks are  
333 considered. This response is also expected if certain risks (e.g. price and yield  
334 volatility) or farmers' risk aversion increase. In addition, the number of animals will  
335 be reduced in order to ensure the availability of fodder to feed all livestock. Since  
336 extensive use of meadows is rewarded by the government by paying subsidies,  
337 this extensification leads to an increasing share of subsidies in the total income of  
338 farmers. Subsidies are paid irrespective of the realized yield of grasslands and  
339 therefore these extensive meadows provide a safe income. Thus, direct payments  
340 can replace other risk management instruments (Finger and Lehmann, 2012).

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## 343 **4 DISCUSSION**

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345 The consideration of yield uncertainty leads to a change in agricultural production  
346 and subsequent changes in land use compared to solutions with no risk aversion.  
347 In this case farmers rely more on the continuous payment of subsidies than on  
348 uncertain yields. Furthermore Bocqueho and Jacquet (2010) show that perennial  
349 energy crops can be a land use option if instead of the Net Present Value (NPV)  
350 the expected utility is maximized. In this case energy crops provide an option to  
351 diversify income and therefore lower the variability of income. This is the case even  
352 if expected NPV of energy crops is lower than that of traditional crops. Since the  
353 gap between the expected NPV of agricultural and forest land use, that can be  
354 expected to have same impact on income than the cultivation of perennial energy  
355 crops, under today's economic conditions still is significant, the transformation into  
356 forests is not yet a considerable risk-management strategy. Under a future open  
357 market scenario this could be changing and forests could, at least on less suitable  
358 parcels, be a strategy to reduce income variability.

359 With respect to the provision of ES, the expected decrease in the number of  
360 animals will lead to a decrease in food provision but an improvement in the carbon  
361 balance. Briner et al. (2012) showed a clear trade-off between food provision and  
362 carbon sequestration. It can therefore be expected that higher risk aversion of  
363 farmers will have a positive impact on carbon sequestration. In addition the risk  
364 aversion and the subsequent increase in extensively used meadows can be  
365 assumed to have a positive impact on biodiversity in this region. Knop et al. (2006)  
366 showed that extensively used meadows are especially rich in species. In addition  
367 under future scenarios the reforestation of certain parcels will increase landscape  
368 diversity, which is also considered positive for biodiversity. Graveline et al. (2012)  
369 also show that the consideration of uncertainty in prices and yields leads to  
370 decisions that reduce the emission of nitrate in two French regions, a trend that  
371 could also be visible in Davos. With respect to the applied agents the impact of  
372 uncertainty on ES provision is expected to be higher if the agents work fulltime on  
373 the farm since then they do not have an additional income that could allow them to  
374 endure years with unfavourable weather or market conditions. For these agents  
375 higher uncertainty is expected to lead to less intensive land use and subsequently  
376 to higher provision of biodiversity and improved carbon balance but less provision  
377 of food.

378 The expected impact of uncertainty however might be slightly overestimated in our  
379 model. The only available risk management options are a change in land use as  
380 well as in the number of animals. For example Mosnier et al. (2011) or Briner and  
381 Finger (2012) show that there are additional risk management options available to  
382 reduce the exposure of farmers to income risk, such as the storage of fodder or an  
383 adaptation of the animals' diet. Our model also does not take costs associated with  
384 variability in the herd size into account. If the size of the herd has to be reduced,  
385 animals can normally only be sold at a lower price than if these animals would be  
386 purchased on the market, causing a net loss. The consideration of these costs in  
387 the model would increase costs associated with risk management. It would  
388 however also be necessary to introduce more but shorter decision periods since  
389 the consideration of these details only makes sense if the agents have the  
390 possibility to adapt their herds during the year, i.e. that they have more flexibility to  
391 adapt to new circumstances

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