Ex ante evaluation of the benefits and costs of agrienvironment measures: An integrated approach based on ecological scorecards and microsimulation

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Introduction

- Agricultural intensification (increased input use), land fragmentation and abandonment → biodiversity loss, pollution, landscape homogenization
- Development of incentive-based policy instruments e.g. agri-environment schemes (AESs) and measures
- → Need for cost- and environmentally effective interventions



Research question

Which measures have the highest potential to enhance the quality of semi-natural grassland habitats at lowest cost?

- → We aim to estimate benefit-cost ratios (biodiversity benefit per € spent)
- → Methodological challenges: Lack of habitat quality and cost data

Agri-environment measures (AEMs)

1	Reduced stock (by 10%)
2	Reduced fertiliser use (by 10%)
3	Reduced crop protection (by 10%)
4	Reduced purchased seed (by 10%)
5	Flail cutting (release of 1 hectare of land)



Cost-effectiveness analysis (CEA)

$$BB_{im} = \frac{\Delta Score_{im}}{\Delta Cost_{im}} = \frac{S_{im (after)} - S_{im (baseline)}}{C_{im (after)} - C_{im (baseline)}}$$

where:

BB: Biodiversity benefit (per hectare) of adopting measure m by farm type i;

 Δ S: Difference in ith farm's biodiversity score between the after adoption scenario of measure m and the baseline scenario (before adoption); Δ R: Difference in ith farm's (negative) FFI between the after adoption scenario of measure m and the baseline scenario (before adoption)



Modelling farms: Environmental data

- Environmental performance data:
 - FARM_ECOS grassland habitat survey derived by using ecological scorecards
 - 8 farms
 - Three levels of stocking rates: High [> 1.4]; medium; low [< 0.7]
 - Three levels of average grassland scores (high, medium, low)



Assessment of habitat quality: FarmEcos scorecard

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Modelling farms: Economic data

- Nationally-representative farm data (Teagasc National Farm Survey):
- → Teagasc NFS (National Farm Survey) records data related to AEMs
 - → Also cost and revenue data from similar farms

Uncertainty analysis: Monte Carlo simulations (1,000 reps)



Methodology: Modelling ecological impacts

 Impacts of measures on individual score components are based on expert opinion



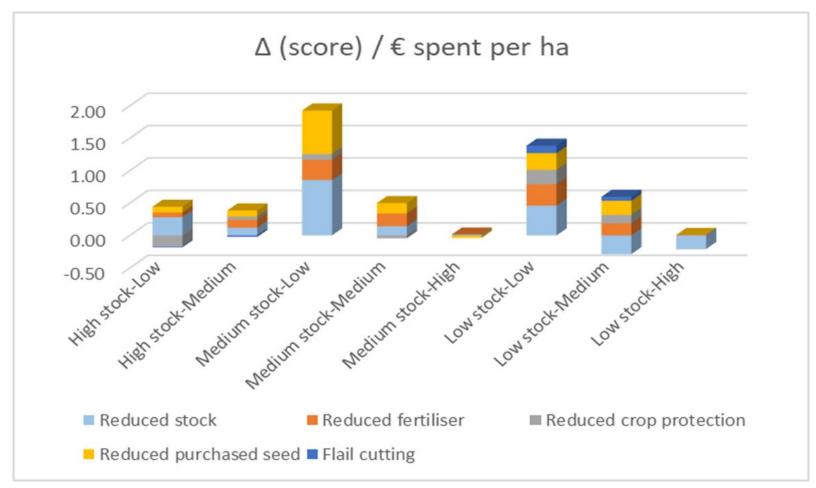
Modelling economic impacts

 Use of Teagasc NFS panel data (2013-17) and separate random effects models for each measure

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$$ln(Cost) = \beta_0 + \beta_1 ln(AEM) + \beta_2 D_Year + \beta_3 D_Region + \beta_4 D_Soil$$



Results: Estimated benefit – cost ratios





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	Reduced stock	Reduced fertiliser	Reduced crop protection	Reduced purchased seed	Flail cutting
	Neduced Stock	Neduced Tertiliser	protection	parchasea seea	i iaii cutting
High stock-Low	0.279	0.072	-0.173	0.090	-0.010
High stock-Medium	0.117	0.124	0.048	0.094	-0.022
Medium stock-Low	0.854	0.311	0.088	0.671	
Medium stock-Medium	0.140	0.197	-0.044	0.160	
Medium stock-High	0.012	0.000	-0.003	-0.035	
Low stock-Low	0.458	0.330	0.220	0.259	0.115
Low stock-Medium	-0.288	0.188	0.127	0.217	0.061
Low stock-High	-0.204	0.007	-0.004	0.000	-0.002



Uncertainty analysis: Monte Carlo simulations

	Estimated mean	Simulated mean	Standard deviation	[95% Confi	dence Interval]
Reduced stock	0.171	0.272	0.335	0.252	0.293
Reduced fertiliser	0.153	0.165	0.095	0.159	0.171
Reduced crop protection	0.033	0.021	0.114	0.014	0.028
Reduced purchased seed	0.182	0.334	0.201	0.321	0.346
Flail cutting	0.014	0.045	0.040	0.043	0.048



Conclusions

- Although the cost-effectiveness of conservation measures varies across farms:
- → Medium stock-Low score and Low stock-Low score farms seem to be more benefited by the measures
 - → Flail cutting is the least effective measure
- Implications for farmers: Assist farmers to make informed conservation adoption decisions
- Policy implications: Assess the cost-effectiveness of payments (e.g. REAP payments)



Acknowledgements







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