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Challenges for integrated assessment and Cost-Effectiveness analysis of mitigation measures for controlling water pollution

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ABSTRACT

Identification of challenges for integrated assessment of mitigation measures proposed to improve the ecological and chemical status of water bodies is explored in this paper. These challenges are addressed within a framework proposed for a spatially-distributed cost and effectiveness analysis. Resulting environmental and economic impacts of the implementation of measures are assessed using agro-hydrological and bio-economic modelling with effectiveness and costs calculated at each sub basin level. Cost-effectiveness ratios calculated over a long period of measures implementation depend on the location where measures are applied. Mapping costs and effectiveness at the basin sub-level reveals the locations where implementing a measure (within interval confidence) would be the most cost-effective. To compare measures between themselves, cost-effectiveness ratios can also be calculated for the whole catchment by adding up sub basin total costs, and by assessing effectiveness at the watershed outlet. Advantages and shortcoming of the integrated approach are discussed with challenges to investigate all the uncertainties related to mitigation programmes of agro environmental policies for reducing water pollution.

KEY WORDS: Integrated assessment, Cost Effectiveness analysis, Mitigation measures, Watersheds.

INTRODUCTION

When formulating programmes of measures to be implemented in the European river basins, the EU water Framework Directive proposed the Cost-Effectiveness Analysis (CEA) as an economic tool for minimizing costs, effectiveness being supposed to be implied. The approach to CEA is however not specified and no standardized approach to CEA is proposed. The aim of this paper is to discuss and evaluate a pragmatic quantitative approach based on a spatially distributed Cost-Effectiveness Analysis (sCEA) applied on two case studies within the Garonne river basin (Upstream Gers river basin) and the Charente river basin (Ne watershed) in terms of their applicability, transparency and capability for decision-making. Conclusions are drawn with regard to the suitability of the approaches and the uncertainties.

Our approach detailed in Lescot (2013) is integrated because of its interdisciplinary nature implying coherence and consistency of the linkages between models. It could be categorized as a bottom-up approach focusing on the cost of implementing measures at the micro level (hydrological unit) and the macro level (watershed) while assessing environmental effectiveness at the sub and watershed level (Fig.1). The effectiveness of some remediation-mitigation measures can be assessed with regard to intermediate goals (reduction of pressure) by the use of indicators and/or final goals (impacts) where models are needed. Given the objectives laid down in the WFD (water bodies to be returned to "good" ecological status by 2015), impacts need to be evaluated. In addition, pressure indicators¹ for pesticides although widely used in policy analysis could back up model outcomes only if they are spatially distributed.

Questions of spatial, temporal, and technological heterogeneity make devising environmental mitigation

programmes a complicated exercise since the characteristics of agricultural production such as soil types, slopes, farming systems or proximity to streams can vary hugely across a river basin. Furthermore, the precise extent of the damage caused to the environment by the use of pesticides is difficult to assess, due to the delay between their application and the appearance of any quantifiable effects. In addition, these relationships are subject to a number of stochastic influences outside the farmers' control.

Finding the right scale has been identified as one of the major challenges, conceptually and methodologically in all science that uses geographic information. For assessing mitigation measures, the scale to select is neither straightforward nor neutral. Nonetheless, to carry out a CE analysis, costs and effectiveness should best be calculated at a common scale appropriate to represent underlying physical processes. Because the watershed or sub basin level is the suitable scale for assessing effectiveness in term of pollutant concentration reduction in outlets, the costs should be assessed at the same scale which may cause difficulties because of the lack of information on farms plots location and practices. In addition to space scale, there should be coherence and consistency for time scale and the choice of the appropriate time horizon. Hydrological modelling offers a means of assessing impacts over a long period of time covering several years of implementation of measures. Beyond that, it is generally advocated that results obtained by modelling are an objective source of information that can be used to support decisions.

To assess the impacts of mitigation strategies, we use the SWAT model (Neitsch et al., 2005) with ArcSWAT². The SWAT model is suitable for predicting long-term impacts of mitigation measures on agricultural chemical yield and for simulating agricultural management practices. The overarching objective of this paper is to address challenges

¹ Such as EIQ (environmental Impact Quotient) or QSA (Quantity of active substances sold), NODU (Number of Used Doses) and IFT (Treatment frequency index) used in France.

² Geographic Information System ArcGIS10 interfaced with SWAT 2009.

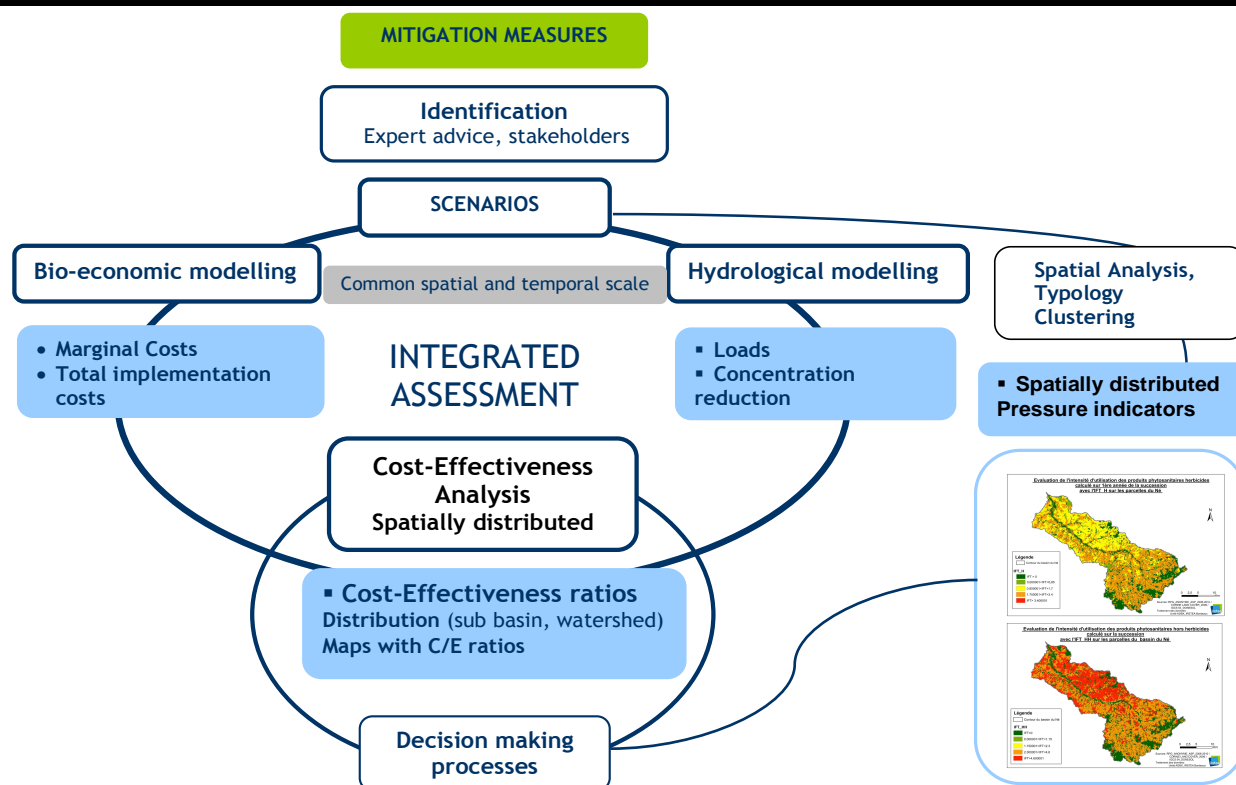


Figure 1. Integrated assessment developed for spatially distributed cost-effectiveness analysis.

in integrated assessment of mitigation measures related to the following elements (Fig.1):

- The economic modelling to assess marginal costs of implementing mitigation measures at the hydrologic unit (HRU)³ and quantify total costs at the catchment level over a period similar with the hydrological simulation period,
- The hydrological modelling to assess over a long period the pollution reduction of pollutants in water streams following implementation of measures,
- The defining of scenarios: Spatial Analysis, Clustering, typologies to assess standard crop rotations and practices and to help build spatially distributed pressure or combined pressure/vulnerability indicators,
- Interaction with stakeholders, communities and governance systems.

CHALLENGES

Limitations of approaches for costs calculation

When choosing mitigation measures, their effectiveness is supposed to be implied. However, if we assume that two measures could have the same effectiveness in reducing pollution, this can be theoretically obtained at different cost

because marginal costs for implementing measures are not equal between themselves, given a location, and for a same measure between locations where it is implemented). Farm models allows traditionally for modelling behaviour of individual farms but requires wide information that is often not available. Using representative farms or type farms is an attempt to overcome these difficulties. Nonetheless, by doing so, we loose geographic information on farms plots and localization of practices. In addition, technological heterogeneity is not adressed with representative (average) farms and not completely with type (modal) farms.

In many studies, the loss in total gross margin is calculated either by a linear programming (LP) model either by partial budgeting. LP models usually (e.g Volk, 2008), optimise agricultural production programmes under different management scenarios and costs are assessed by comparison of gross margins. On the other side, econometric approaches can help assess the estimated changes in the gross profit margins of a large data set of farms (e.g., Fezzi 2001). Moreover, econometric models automatically provide estimates of uncertainty derived directly from statistical inference rather than requiring sensitivity analysis. Nonetheless we preferred the methodology of economic optimisation, more relevant when dealing with limited and incomplete information and able to link economic elements with ecological and biophysical elements.

The alternative form of aggregation we used, which overcomes above scale problems, combines farm plots in each spatial unit together, treating them as a single entity ensuring consistency with the environmental modelling.

³ Hydrological Response Units (HRU), consisting of homogeneous land use, management and soil characteristics, and therefore different in size. HRUs are the spatial units used by the SWAT model to calculate the vertical, lateral and sub-lateral flows of water and nutrients. These flows are then aggregated for each sub basin. Water and pesticides from HRUs in sub watersheds are routed to the sub watershed outlets.

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entry for the calculation of spatial environmental indicators (a part of the integrated assessment process not detailed in this paper). Percentages of area for each crop and each type of soils within the watershed limits are validated afterwards by expertise. Files generated by this model are able to be directly used by the SWAT model. Because costs are calculated on a one-year basis with the bio-economic model, the shifting from rotations to land use is carried out by allocating land of the HRU to crops proportionally to their part in the crops sequence over the rotation period.

Limitations of approaches for hydrological modelling

Once the SWAT model is updated and calibrated, it can be used to simulate the pesticide mitigation strategies and evaluate their effectiveness. The effectiveness is presented in terms of absolute reduction in pesticide concentrations (in $\mu\text{g.l}^{-1}$) following the implementation of a measure. The first step of the model is to subdivide the watershed into sub basins that are further disaggregated into Hydrological Response Units (HRU). The management practices are defined at the HRU level by specific management operations (beginning and end of growing season, timing of cultural operations, amount of fertilizers and pesticides, irrigation management). Loads then concentrations of nutrients and pesticides are calculated at the outlet of the whole watershed, and of each sub basin from the SWAT main channel output files with reference concentration of the baseline scenario. To consider the problem of pesticide time lags (Roa-García and Weiler, 2010), simulations are performed over a 25-year period with measures being applied each year. For the final pesticide concentration, we retained the average value of pesticide concentrations calculated over the last ten years of the modelling period.

Concentration reduction is calculated as the difference between these average concentrations:

$$([C_0] - [C_s]) \quad (2)$$

$[C_s]$: average concentration over the ten last years of hydrological simulation ($\mu\text{g.l}^{-1}$) with measure applied each year (for a given area); $[C_0]$: baseline concentration ($\mu\text{g.l}^{-1}$).

For water flows, predicted values generally matched well with the observed values for both calibration and validation periods (coefficient of determination and Nash-Sutcliffe Efficiency index). Uncertainties on the irrigation water volumes actually applied by farmers and the lack of detailed information on the management of low water regimes by water agencies may sometimes explain discrepancies between simulated and observed daily values.

Modeling pesticide run-offs is more complicated because of the number of active substances and diverse protection strategies. In our case studies, we restricted the number of active substances to the most widely used molecules and/or the molecules the most frequently found in sampling water analysis. The set of active substances⁶ applied to each main crop of the watershed is listed with the frequency of their use either from expertise (Ne watershed) either from surveys carried out on farms with additional expertise

(Upstream Gers river basin). In this late case, we defined new “average active ingredients” (AAI) based on physical and chemical properties⁷ (Koc, DT50 and solubility) of the molecules for each pesticides group with the average dose (arithmetic mean) applied by farmers. In the Ne watershed study, each of the ten widely used molecules have been simulated separately. Average simulated concentrations and measures occasionally differ significantly and these differences between simulated and measured concentrations are difficult to handle because of the uncertainty of the measurements themselves (related to their low frequency and the limited number of values within a month and year).

Limitation in the use of Cost-Effectiveness ratios

CEA is an appropriate approach for evaluating mitigation measures where the measurement of benefits is difficult and unsure. CEA cannot determine the overall value of a single measure but is particularly useful in comparing two or more measures. It summarizes results into single useful quantitative indicators for selecting measures. For the sCEA, we used a spatialized cost-effectiveness indicator R, defined at the sub basin or watershed level, given the area implemented with the measure. Thus, for a given area implemented with measures we have:

$$R_{\text{subbasin/basin}} = \frac{\text{discounted sum of annual costs (€)}}{\text{concentration reduction (} \mu\text{g.l}^{-1} \text{)}} \quad (1)$$

This ratio enables measures to be ranked in terms of increasing unit costs per unit of pesticide concentration reduction, given the range of uncertainty. Calculated at the sub basin level and represented on maps, they allow for rational discussion between stakeholders who often require integrated information. Such maps give an overview of the cost and effectiveness of the various possible mitigation measures, depending on the location where they could be implemented and to which extend WFD objectives could be reached.

MANAGING UNCERTAINTY

A lack of scientific knowledge and empirical evidence are part of the difficulties in the political negotiation process surrounding the formulation of concrete water policy objectives, including indicator parameters and specific threshold values⁸ (Brouwer and Blois, 2008). Uncertainties can be found at each step of our analysis. There is first uncertainty relating primarily to the identification of the environmental objectives (restoration of good ecological status) and their monitoring (EU health-based drinking water pollutant concentration limits and method detection limits). There is next uncertainty relating to the identification of the main pollution sources (point and non-point) and of their

⁶ All pesticides that are officially registered for use and are likely to be used in the watershed area should be monitored and the total sum calculated by adding all concentrations that exceed the parametric value/detection limit of 0.1 $\mu\text{g.l}^{-1}$. Thus there is no standard about the number of pesticides that should be considered in total as this will vary between watershed areas and could add up to hundreds of pesticides.

⁷ Koc: Sorption coefficient normalized to organic carbon content; DT50: Time for a 50% decline of the initial pesticide concentration

⁸ The value of 0.1 $\mu\text{g.l}^{-1}$ in the WFD is a substitute for zero i.e. absence in water or below the detection limit. The parametric values for individual pesticides (0.1 $\mu\text{g.l}^{-1}$) and for total pesticides (0.5 $\mu\text{g.l}^{-1}$) are not based on any scientific findings. Besides, the World Health Organization uses a different approach with different set of guideline values for a large number of individual pesticides.

contribution to the water quality problem (agriculture, gardening, ditches or railways maintenance).

Point-source pollution by pesticides mainly relates to pesticides handling (filling, cleaning, remnant management) while the non-point pollution sources are more related to the application of pesticides and often result from natural environmental factors (runoff, drift etc). Few studies are available on the subject, in which point and non-point sources are clearly separated. In the literature, it is estimated that point sources could contribute up to 50% to the pollution of surface water by pesticides (Leu et al., 2004; Bach et al., 2005). Moreover average practices do not take account of excessive use of pesticides.

There is uncertainty also relating to the identification of potential mitigation measures proposed by science to solve the water-quality problem. Effectiveness is firstly tested on the plot scale and only assessed by modelling on a large scale). Moreover, best location for implementing the measures is not straightforward as the zones identified by the pollution pressure may not contribute the most to pollution at the outlet. Finally there are uncertainties in the modelling itself due to the availability of data (limited water analysis and not always at appropriate times of flow for calibration and validation processes) and uncertainty on values. Sensitivity analysis is usually carried out when using the SWAT model for selected parameters and different management practices. Classification of existing methods of sensitivity analysis refers to the way parameters are treated: local techniques concentrate on estimating the local impact of a parameter on the model output while global techniques analyse the whole parameter space at once (van Griensven et al. (2006). These authors propose a sampling strategy allowing a global sensitivity analysis with only limited number of model runs overcoming the problem of over parameterization for distributed model like the SWAT model.

Uncertainties and error margins in the estimation of the costs should be considered as well with bio economic modelling. In sensitivity analysis, consideration is given to the way in which errors in a set of input data affect the error in the final output. In practice we could suppose that varying the number of constraints will have no effect on the costs calculation. On the contrary, varying the number of activities by proposing other new crops with environmentally friendly practices may change the costs calculated. Much attention should be paid also to the change of the objective function (minimizing risk instead of maximizing profit), the values for sub optimal strategies or the variation of uncertain technical parameters (such as prices of inputs and outputs) that would help to observe changes in calculated costs of mitigation measures. Because the bio economic model is used for comparison of costs and not for their absolute values, we could suppose that errors in input data affect the costs calculation the same way for the different measures.

Questions arise lastly with the use of the CE indicator on its level of confidence as sensitivity analysis both for hydrologic and economic/economical assessments will lead to a range of values determined by the way how parameters will be treated.

CONCLUSION

J. D Brown (2004) argues in favour of explicit assessments of uncertainty in environmental data and models as a condition for balancing uncertain scientific arguments against uncertain social, ethical, moral and legal arguments

in managing environmental systems. A core challenge lies in minimizing all these uncertainties. The cost-effectiveness framework presented was developed to be used as a pragmatic support tool for policies analysis envisaged here in terms of alternative allocations of resources, the objective being to find the agro environmental measures that contribute most to achieving goals at minimum costs. Such an approach could help Basin Committees or Water Agencies better target implementation of measures and financial incentives to farmers where appropriate. A lack of cost-effectiveness is one of the major drawbacks of incentive schemes for mitigation measures and their availability is also restricted by a lack of funds. Based on CE approaches, the choice of a mitigation programme could be then a system of locally specified management incentives.

LITERATURE CITED

Paper in journal

- Bach M., Röpke B., Frede H-G. 2005. Pesticides in rivers – Assessment of source apportionment in the context of WFD. *European Water Management Online*. EWA, 14p
- Brown, J.D., 2004. Knowledge, uncertainty and physical geography: towards the development of methodologies for questioning belief, *Transactions of the Institute of British Geographers* 29 (2004), pp. 367–381.
- Brouwer R. & De Blois C., 2008. "Integrated modelling of risk and uncertainty underlying the cost and effectiveness of water quality measures. *Environmental Modelling & Software* 23(7): 922-937.
- Fezzi C. & Bateman I.J. 2011. Structural agricultural land use modeling for spatial agro-environmental policy analysis, *American Journal of Agricultural Economics*, vol. 93, pp. 1168-1188.
- Lescot J.-M., et al. 2013 A spatially-distributed cost-effectiveness analysis framework for controlling water pollution, *Environmental Modelling & Software* 41(0): 107-122.
- Leu C., et al., 2004. Simultaneous assessment of sources, processes, and factors influencing herbicide losses to surface waters in a small agricultural catchment. *Environ. Sci. Technol.*, 38, 3827-3834.
- Roa-García, M. C. & Weiler M. (2010). "Integrated response and transit time distributions of watersheds by combining hydrograph separation and long-term transit time modeling." *Hydrol. Earth Syst. Sci. Discuss.* 7(1): 1-32.
- van Griensven, A., et al. (2006). "A global sensitivity analysis tool for the parameters of multi-variable catchment models." *Journal of Hydrology* 324(1-4): 10-23.
- Volk M., et al. 2008. Integrated ecological-economic modelling of water pollution abatement management options in the Upper Ems River Basin. *Ecological Economics*, vol. 66: 66-76.

Paper in Proceedings

- Vernier F. et al., (2010). Modelling scenarios of agriculture changes on freshwater uses and water quality at a large watershed scale: the case of the Charente watershed (France); 2010 International Congress on Environmental Modelling and Software, 5th Biennial Meeting, July 5–8 2010, Ottawa, Canada

Book

- Brooke A., Kendrick D., Meeraus A., 1988. GAMS: A User's Guide. The Scientific Press, South. San Francisco, CA, 1988.
- Mc Carl B.A., et al. GAMS User Guide; 2009.Version 23.3. GAMS Development Corporation.
- Neitsch S.L., et al., 2005. Soil and Water Assessment Tool Theoretical Documentation, version 2005. Temple, TX: Grassland, Soil and Water Research Laboratory, Agricultural Research Service.