

# The treatment of public participation GIS data in environmental management: setting up smart systems for the synthesis and mapping of vague definitions

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## 1. Introduction

This report describes a research protocol and software implementation for the synthesis, retrieval and mapping of georeferenced documents. The model is based on the addition of written text to events located in space, laying theoretical foundations for Information Systems with both geospatial and text mining functionality. We focus on human languages and describe a routine for the treatment of written text provided from spatial surveys, in the context of crowd-sourcing environmental information.

Fig. 1 depicts the outline of the research protocol. We consider Stages 1 and 5 to be of universal application in studies founded on principles of the scientific method, whereas Stages 2 and 3 have already been treated before (e.g. Carver *et al.*, 2009; Gunderson, 2006; Gunderson and Watson, 2007). The paper hence centers on the technical basis for the execution of Stage 4, demonstrating the full research process, however, with a study on public perceptions of environmental change on the Flathead Indian Reservation (Montana).

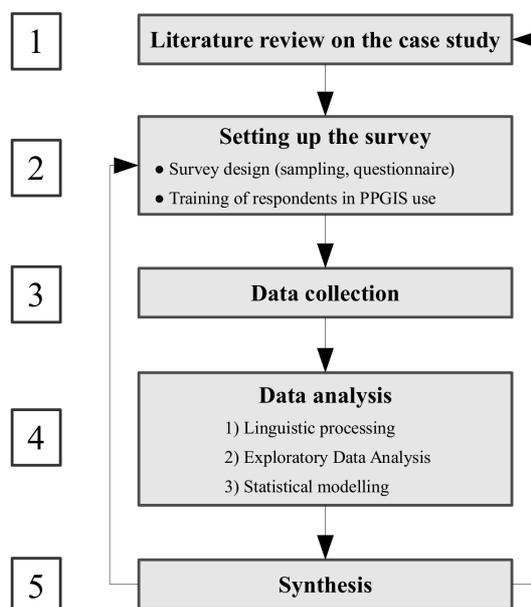


Fig. 1. Workflow followed in this study.

## 2. The role of Participatory GIS in environmental management

Changes in local ecosystems, as documented throughout the planet over the last decades, have been associated with a steep rising of the Earth's average surface temperature on a global scale (Solomon *et al.*, 2007). Deciding what strategies to take so that habitats and human communities are resilient to these impacts has hence become a significant challenge, in the context of which risk and monitoring have taken a major role in setting the pace of policy and planning agendas (e.g. the UN's Millennium Development Goals<sup>1</sup> or the EU's Europe 2020<sup>2</sup> strategy; see also Field *et al.*, 2012; Parry *et al.*, 2007; Metz *et al.*, 2007). In this respect we believe in the positive effects of including the social actors, their interests and perceptions in the process of policy-making, planning and implementation, given the capacity of public engagement to catalyze the consolidation of decisions taken at the societal scale (Pickles, 1995). Participatory Geographic Information Systems (PGIS) have the capacity to become instrumental in documenting such interests and perceptions, by providing technologies for the involvement of communities in the description of space (Kingston, 2007). When coupled with textual data, PGIS can additionally allow the segregation of landscape

1 <http://www.un.org/millenniumgoals>. [Accessed on 17/12/2013]

2 [http://ec.europa.eu/europe2020/index\\_en.htm](http://ec.europa.eu/europe2020/index_en.htm). [Accessed on 17/12/2013]

properties on the basis of the meanings people ascribe to locations, and lead to a better understanding of spatial relationships between tangible and intangible elements of humanized ecosystems (Carver *et al.*, 2001). For instance, PGIS platforms have been used to describe individual perceptions of wilderness among traditional communities (Carver *et al.*, 2009).

### **3. The problem of vagueness in the qualification of data**

A major impediment to the retrieval of information from textual data is the allocation of meaning to vague descriptions, and the uncertain definitions that derive from it. Uncertainty *sensu lato* is a property of decision-making processes that arises from imperfect knowledge about the initial formation and subsequent development of systems. The notion of vagueness can thus be placed within a general taxonomy of Uncertainty where specific properties of qualitative and quantitative data types are distinguished (Table 1). Following this ontology, the report considers the effects of value ambiguity, or more specifically of semantic ambiguity as it can be understood after a closer consideration of the concept (Table 2), on taking account of qualitative properties in textual data.

### **4. Solutions to information retrieval from the automated processing of human languages**

Retrieval of semantic information out of textual data is implemented by Natural Language Processing (NLP), a main component of Artificial Intelligence where computer science and linguistics converge for the development of verbalized human–computer interaction systems (Chowdhury, 2003; Joshi, 1991). NLP makes use of machine learning, data mining, computational linguistics and, more broadly, is founded on principles of statistical linguistics and cognitive linguistics. Synthesis and retrieval of semantic properties out of textual data hence involves an analytical problem where application of NLP-based procedures naturally emerges, especially of those aiming at morphological segmentation (e.g. for the generalization of words as neutral forms), named entity recognition, word sense disambiguation, co-reference resolution (e.g. anaphora resolution), part-of-speech tagging, sentence breaking, syntactic analysis, text simplification, automatic summarization, natural language understanding (e.g. through first-order logic), sentiment analysis and topic segmentation (Jurafsky and Martin, 2008; Manning and Schütze, 1999).

**Table 1. A taxonomy of uncertainty.**

Type of uncertainty	Explanation	Type of data	
		Quantitative	Qualitative
Classification ambiguity	Variable systems to allocate values to classes.	•	•
Value ambiguity	Feasibility that several values may be allocated to a single event.	•	•
Accuracy	Measurable difference between the observed value and the real one.	• (Metric)	• (Topological)
Completeness	Occurrence of missing data.	•	•

**Table 2. A taxonomy of value ambiguity.**

Type of ambivalence	Explanation	Type of data	
		Quantitative	Qualitative
Decimal redefinition	Re-expansion of the decimal digits of a previously rounded or truncated quantity due to numerical generalization.	•	
Semantic ambiguity	Interpretation of a data unit with alternative meanings.		
a) Subclass redefinition	Redefinition of a superclass term with one of its subclasses.		•
b) Homonymy	Redefinition of a term having several meanings (e.g. pronoun precision and polysemy).		•
c) Semantic inference	Alternative meanings can be inferred from a text when testing textual entailments.		•
Random ambivalence (aka numerical precision)	The value of an event is modelled as a stochastic realization of a probability model. The dispersion of feasible values gives the precision of the measurement.	•	•
Procedural ambivalence	Different results are obtained on applying alternative, suitable methods.	•	•
Positional ambiguity (aka positional vagueness)	An event may be located at different positions across a delimited spatial domain.		
a) Inclusive	All positions are valid.	•	
b) Exclusive	The actual position invalidates the possibility that others occur.	•	
State ambiguity (aka state vagueness)	An event may have had different states across a delimited time domain.		
a) Inclusive	All states are valid.	•	•
b) Exclusive	The actual state of the event invalidates the possibility that others have occurred.	•	•

## 5. Setting up intelligent geographic information systems: the multimodal geodatabase

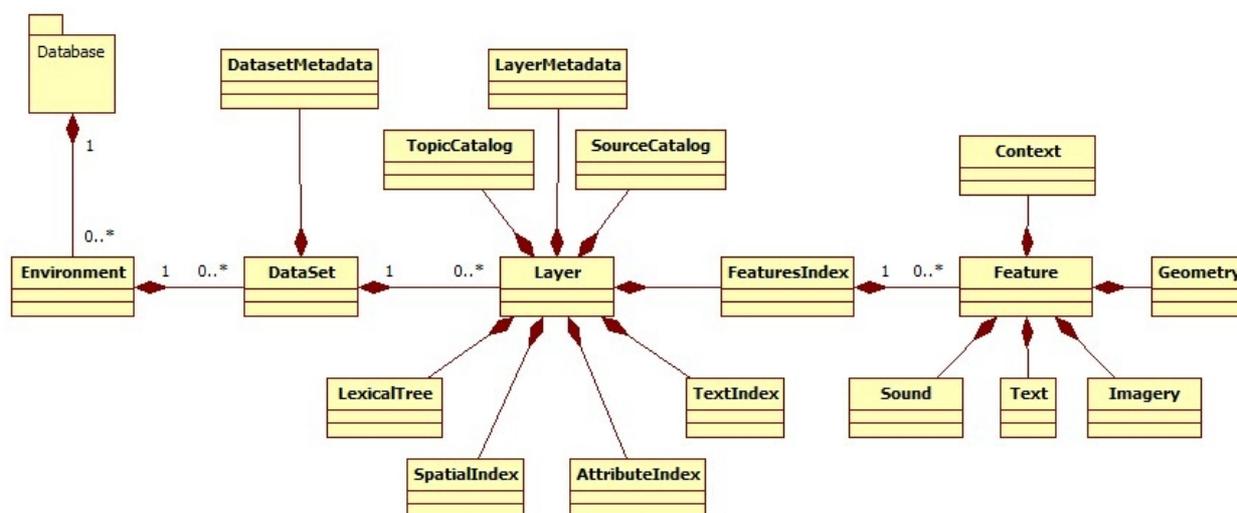
In the present study we apply the concept of *multimodality* (Jewit, 2009; Kress, 2010) to refer to spatialized objects that, potentially, may contain semantic information in the form of co-located context (i.e. references to an external geospatial process that coincide with the object in space-time), text, imagery and audio. As shown in Fig. 1, Stage 5 is actually a three-step process that (1) starts with linguistic synthesis and organization of textual data (e.g. by the summary and clustering of linguistic properties), (2) proceeds to Exploratory Data Analysis for description of multidimensional patterns in the multimodal geodataset and (3) concludes with the statistical modelling of such patterns, paying special attention to the testing of spatial processes explaining the distribution of verbalized information. Ultimately, the analysis aims at the retrieval of meanings provided by data sources, allowing the mapping of human perceptions and the testing of associations between messages and the spatiotemporal context where such messages were produced.

Central instruments in this analysis are (i) a logical data structure for the storage of multimodal data and (ii) an interface that enables their retrieval and examination. In our current data ontology each object in the multimodal geodatabase is a geospatial *feature* that may contain both linguistic and measure-based attributes. For multimodal data storage we thus consider an Object-Oriented geodatabase where events may have one or some of the above types and related attributes – e.g. geometric (spatial coordinates, geometry type), co-locational (measures on external phenomena), textual (lexicon, syntax, sentiment, textual entailments). Time is expressed by grouping all contemporaneous features into the same *time* layer of a multimodal geodataset (Fig. 2).

Semantic retrieval and mapping is done by an *ad hoc* extension of Structured Query Language that implements NLP functionality through a naturalized query language, in keeping with the Natural Language Programming paradigm (Veres, 2008). The present development aims at testing implicit meanings, e.g. by the syntax `select [those] features where [their] text entails` followed by the statement to be evaluated, as well as the assessment of semantic vagueness by commands such as `select [those] features where [their] text has [the] noun heather or [its] superclass` (words between brackets can be omitted), based on conceptual hierarchies like those of the WordNet corpus<sup>3</sup>.

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3 <http://wordnet.princeton.edu>. [Accessed on 18/12/2013]



**Fig. 2. Class diagram of the integral database.** All associations between classes are One-to-One compositions of the 'has a' type (i.e. the owner class, the one with the filled diamond attached, has strictly one instance of the owned class) unless otherwise stated.

## 6. Case study: fire ecosystems and traditional ecological knowledge in the northern Rocky Mountains

We show some capabilities of this information system with a PGIS case study on public perceptions of environmental change in fire ecosystems of the Flathead Indian Reservation (Montana, USA) (Fig. 3), within the wider topic of investigating interrelations between science, technology and Traditional Ecological Knowledge (TEK). A major source of concern on the reservation is the effect of fire suppression on the spatial structure and health of tribal woodlands and related habitats. The widespread implementation of federal policies against wildfires over the last century has led to forest ecosystems afflicted by oversized tree communities, anomalous accumulations of dead wood on the forest floor, dense understories of brush and young trees and closed forest canopies (CSKT, 2005), which has largely contributed to an increase in seasonal episodes of uncontrollable wildfire, the reduction of soil moisture, a decrease in sunlight to the forest floor and a proliferation of plant pathogens and disease (CSKT, 2013).

To provide a better depiction on how local communities perceive the current state of tribal woodlands, the Confederated Salish and Kootenai Tribes have closely collaborated with the Aldo Leopold Wilderness Research Institute in a novel approach that seeks for synergies between academia and native communities in natural resource management. Within this framework of tribal

involvement a semi-structured PGIS survey was conducted under tribal supervision among a sample of twenty-nine key informants representing tribal and non-tribal residents of the Jocko landscape unit, with emphasis placed on (i) residents' mapping of the locations of recent changes in their local environment, (ii) residents' attachment of meanings to those locations, and (iii) residents' perceptions about the potential application of TEK to help promote recovery and resilience of those locations.

The survey comprised a demographic questionnaire, a thematic questionnaire on local environmental knowledge, and a "spray and say" questionnaire with fuzzy marking of places and subquestions about meanings attached to those places (Appendix A). Data collection was carried out with the Map-Me ("Mapping Meanings") PGIS tool. A distinguishing feature of Map-Me is its ability to document vague locations of geospatial features by a fuzzy marker (the "spray-can" tool) following the initial approach of Waters and Evans (2003) and Evans and Waters (2007) for raster grids, and more recently Huck *et al.* (2013) for vector models, which enables a suitable means to take account of the vagueness of people's perceptions on expressing spatial features of their environment (Fig. 4).

Twenty-eight respondents provided *spray patterns* in response to the question "Indicate an area whose environmental characteristics you believe have changed in the last decades" (Fig. 5). In order to standardize respondents' behavior on tagging space, a frequency model deriving from overlapping rasterized spray patterns were used instead of the raw multipoint objects. In this model, for each individual spray pattern the algorithm rasterizes the raw multipoint distribution as an *indicator surface*, such that each cell of the surface is given the value of 1 if one or more points occurs in it, and 0 otherwise. Indicator surfaces from all spray responses are next added and a *relative frequency surface* is computed with regard to how frequently every cell is sprayed, by the transformation  $Cell\ value = \frac{Number\ of\ overlapping\ indicator\ surfaces\ marking\ the\ cell}{Total\ number\ of\ spray\ patterns\ in\ the\ sample}$ .

The maximum frequency of overlapping spray patterns in a given cell is 36 % of the total number of spray patterns (Fig. 6). Spots where highest frequencies of respondents agree on having observed environmental changes are found along the Jocko basin and in the east end of the studied landscape unit, which is the southernmost sector of the Mission Mountains Wilderness area. At present, responses are also being processed by the multimodal software implementation here presented, with the goal of grouping and mapping spray patterns according to the semantic

properties of their attached comments. In relation to the query `select features where use of prescribed fire is good`, 31 % of spray patterns support the use of prescribed fire in sprayed spots, with the maximum frequency of overlapping spray patterns being 12 % of all spray patterns (Fig. 7). Sites most frequently marked by non-tribal residents stretch along the Middle Jocko and in herbaceous lands between the basin and higher woodlands, whereas spots most frequently referred by tribal members concentrate in the mountain woodlands and on summits of the wilderness area.

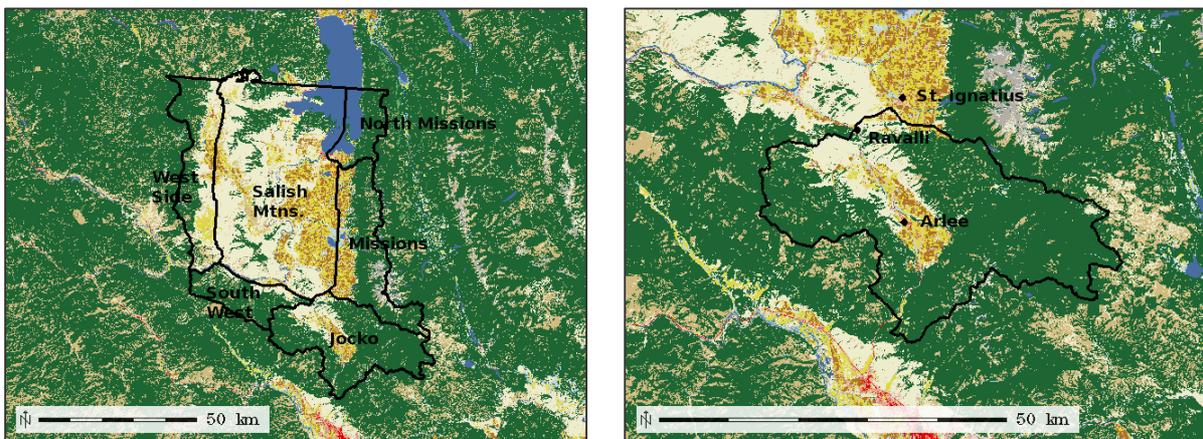
## 7. Discussion and conclusions

An initial conclusion from Fig. 7 is that, when demographic attributes are disregarded, a subsample of both tribal and non-tribal residents agree on a common idea – that restoring prescribed fire in the Jocko landscape unit would have a positive effect on local environments. However, internal differences emerge when geospatial and demographic properties are jointly taken into account. Clearly, opinions differ about where prescribed fire should be introduced. Besides, these seem to be somehow linked to cultural identity based on tribal membership.

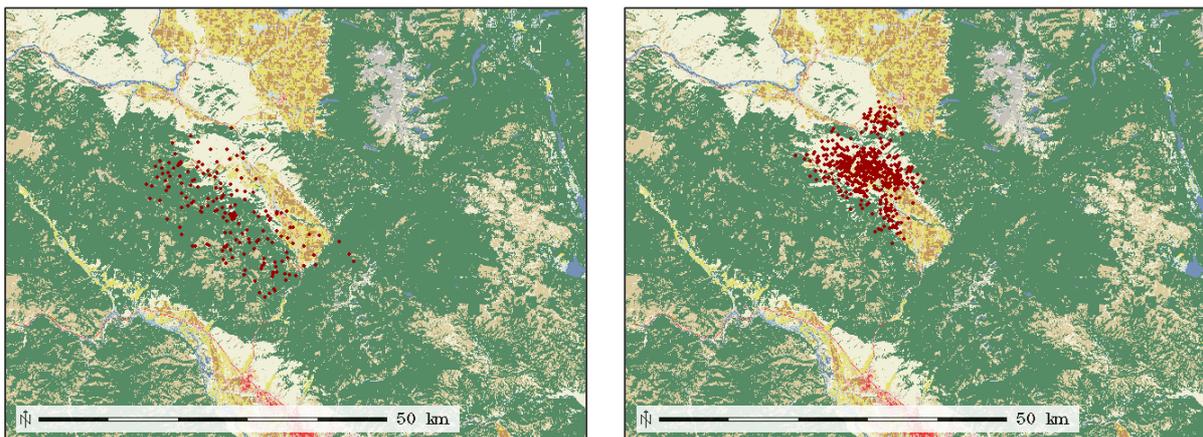
A comprehensive interpretation of this pattern still needs further exploratory analysis, nevertheless, as well as testing of spatial association hypotheses. At least two questions should be examined: (i) what is the probability of the observed associations between spray patterns and geographical location to be the product of chance, and (ii) what landscape properties may explain the observed relative frequencies. Candidate substratum covariates might derive from variables about land use, the evolution of fire regimes and the recent history of land status. In addition, methods of analysis at the inferential level could consist (a) in geosimulation, replicating the use of the spray-can tool by respondents according to some spatial association hypothesis (e.g. testing that spraying behavior relates to specific landscape attributes), (b) in reshuffling the assignment of spray patterns to cultural labels (i.e. to respondents) by resampling techniques, and (c) in the fitting of spray patterns to spatial regression models adapted to spray patterns.

Although it is evident that coupling verbalized expressions with geospatial and demographic properties should enhance information outputs (compared to taking these components separately), our results confirm the perception that treating and analyzing multimodal phenomena does involve a significant leap with regard to reductive perspectives. The “spray and say” approach enables a

more inclusive collection of people's spatial perceptions and ways of expression, but it is the processing of multimodal datasets that allows an holistic treatment of multiple modes and dimensions that objects may take. This processing would further improve with comprehensive multimodal information systems able to couple a large number of modes simultaneously (e.g. verbal, pictorial and measure-based, to name the most usable), clearing the path towards more versatile models of cognition.



**Fig. 3.** The Flathead Reservation (left) and the Jocko Landscape Unit (right).



**Fig. 4.** Use of the fuzzy marker. Each of these two examples of a spray pattern comprises a whole 'coat of spray paint' drawn by a different respondent with regard to a common question, so they each link to a single text object (in this case, a sequence of comments provided by the respondent) that qualifies the full coat.

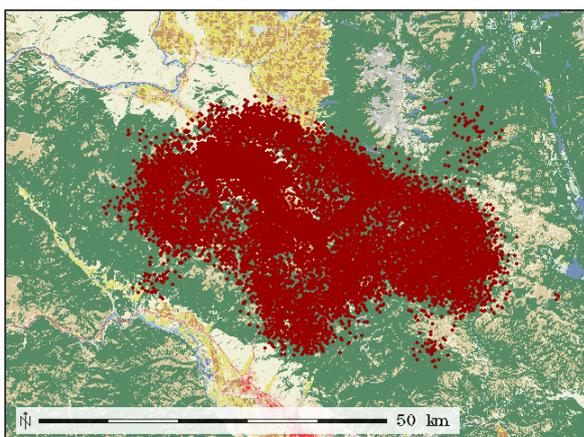


Fig. 5. Joint mapping of all spray patterns marking observed environmental changes.  $N(\text{spray patterns}) = 28$ .  $N(\text{dots}) > 20,000$ .

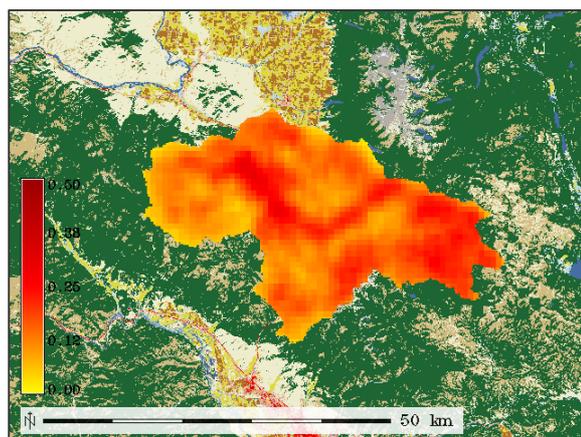


Fig. 6. Heat map of Fig. 5, as the relative frequency that a given location is marked as a spot having undergone environmental changes.

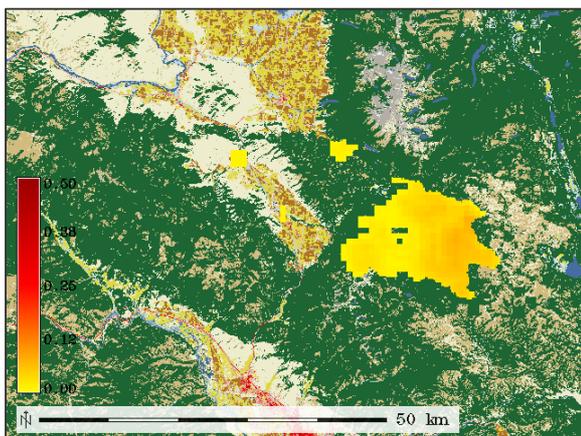
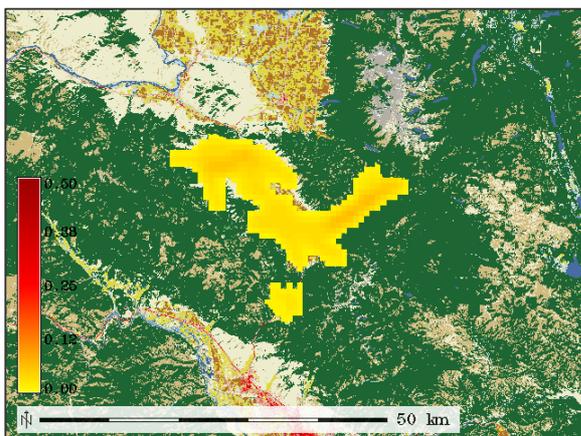
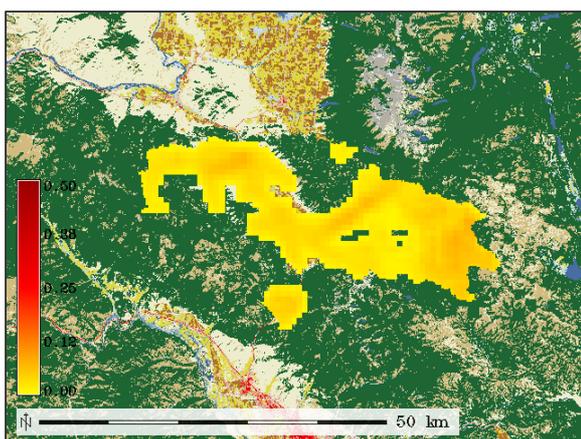
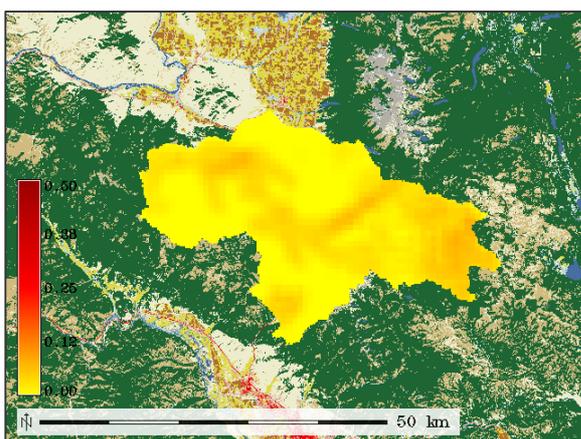


Fig. 7. Relative frequency that marked locations contain comments supporting *in situ* use of prescribed fire. Upper left: all places. Upper right: places where the frequency is higher than 0. Lower left: answers provided by non-tribal residents. Lower right: answers provided by tribal residents.

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## **Appendix A. Questionnaire of the PGIS survey on the Flathead Reservation**

### Demographic questions

- Are you an enrolled member of the Confederated Salish and Kootenai Tribes (CSKT)?
- What is your age group?
- What is your gender?
- What is your occupation?
- If retired, what was your occupation?
- What is your role in Tribal affairs (e.g. official position, committee participation, Elder)?

### Traditional Ecological Knowledge questions

- How familiar are you with the Jocko Landscape?
- How worried are you about climate change?
- How much do you think climate change will affect native plants in the Jocko Landscape (such as from rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel, if at all, that traditional knowledge can help solve these impacts or problems for this area? Please give an example.
- How much do you think climate change will affect wild animals in the Jocko Landscape (such as rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel, if at all, that traditional knowledge can help solve these impacts or problems? Please give an example.

- How much do you think climate change will affect fisheries in the Jocko Landscape (such as rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel, if at all, that traditional knowledge can help solve these impacts or problems? Please give an example.
- How much do you think climate change will affect crops in the Jocko Landscape (such as rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel that, if at all, that traditional knowledge can help solve these impacts or problems? Please give an example.
- How much do you think climate change will affect livestock in the Jocko Landscape (such as rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel, if at all, that traditional knowledge can help solve these impacts or problems? Please give an example.
- How much do you think climate change will affect outdoor recreation places (such as parks, beaches, lakes, rivers, forests, etc.) in the Jocko Landscape (such as rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel, if at all, that traditional knowledge can help solve these impacts or problems? Please give an example.
- How much do you think, if at all, climate change will affect water quality, quantity or location in the Jocko Landscape (such as rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel, if at all, that traditional knowledge can help solve these impacts or problems? Please give an example.
- How much do you think climate change will affect trees, shrubs, and grasses in the Jocko Landscape (such as rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel, if at all, that traditional knowledge can help solve these impacts or problems? Please give an example.

- How much do you think climate change will influence wildland fires in the Jocko Landscape (such as rivers flooding, wildfires, droughts, damage from bark beetles, temperature increases, etc.)?
- How do you feel, if at all, that traditional knowledge can help solve these impacts or problems? Please give an example.
- How well do you feel that these questions have allowed you to express your beliefs about climate change?

“Spray and say” questions

Cartographic question:

- The area outlined in red is the Jocko Landscape. Please indicate an area that you believe has changed over the years.

Subquestions about marked spots:

- What did this area used to be like and what is the source of your knowledge?
- What is the area like now and what do you believe has caused the change from what it used to be like?
- What would you like this area to be like in the future and why?
- What actions need to be taken and what do you believe will be the primary obstacle to achieving this end state?