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Using systematic conservation planning to establish management priorities for freshwater salmon conservation, Matanuska-Susitna Basin, AK, USA

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Abstract:

1- The Alaskan Matanuska-Susitna Basin (MSB) provides habitat for all five Pacific salmon species, and their large seasonal spawning runs are important both ecologically and economically. However, the encroachment of human development through urbanization and extractive industries poses a serious risk to salmon habitat in the MSB.

2- Using systematic conservation planning techniques, different methods of incorporating anthropogenic risks were assessed to determine how to cost-effectively conserve salmon habitat in the area.

3- The consequences of four distinct conservation scenarios were quantified: no consideration of either urbanization or extractive industries (‘Risk ignored’ scenario); accounting for the risk of urbanization, and avoiding conservation in all fossil fuel rich areas (‘Urbanization accounted, all extraction avoided’ scenario); accounting for urbanization and oil and gas development, but avoiding conservation in coal rich areas...
(‘Urbanization accounted, coal areas avoided’ scenario); and accounting for all anthropogenic risks to habitat, and allowing conservation in oil, gas, or coal rich areas (‘All risks accounted’ scenario). To compare conservation success and resiliency, the impact of these risks were estimated using Monte Carlo simulations. The final cost of each solution was then divided by the number of conservation targets met to determine a return on investment.

4- Results from scenarios that avoided all extractive activities, or just coal, suggest that conservation targets cannot be met by simply avoiding fossil fuel rich areas, and these scenarios resulted in lower returns on investment than when risks from extraction were incorporated into the solution.

5- By providing economically rooted conservation prioritization, this study provides a method for local managers and conservation groups to identify conservation opportunities in MSB river basins.

**Keywords:** River, Disturbance, Habitat Management, Landscape, Fish, Industry, Mining, Urban Development
1. Introduction:

Quantifying and incorporating the uncertainty surrounding the potential success of management actions is crucial to making cost effective conservation decisions. A key source of uncertainty is the risk posed to natural ecosystems by anthropogenic activities, a factor that is critical to incorporate in order to give conservation actions the best chance of success (Bode et al., 2009, Tulloch et al., 2013). For landscapes threatened by events that negatively impact biodiversity, quantifying the spatial distribution of risk sources, and including them into conservation plans can increase the overall return on conservation investments (Hammill, Tulloch, Possingham, Strange, & Wilson, 2016). In many parts of the world, landscapes with high biodiversity are threatened by encroaching housing development, as people seek to live near areas of natural beauty. In addition, growing populations increase the demand of natural resources such as oil, gas, and coal. For areas experiencing both population growth and increased pressure on local natural resources, quantitatively assessing where development should and should not take place is crucial to ensure the survival of local ecosystems and their species (Butt et al., 2013).

The Matanuska-Susitna Basin (MSB) covers over 25,000 square miles (approximately 64,750 square kilometres) of south-central Alaska. This basin provides habitat for all five Pacific salmon species (Supplementary material): Chinook salmon (*Oncorhynchus tshawytscha*), chum salmon (*Oncorhynchus kisutch*), coho salmon (*Oncorhynchus keta*), sockeye salmon (*Oncorhynchus nerka*), and pink salmon (*Oncorhynchus gorbuscha*). The ecological importance of salmon spans both aquatic and terrestrial ecosystems. Spawning salmon feed bears, wolves, eagles, and other streamside animals, and after completing their life cycle they provide carbon, nitrogen, and
phosphorus to streams and surrounding riparian areas (Juday, Rich, Kemmerer, & Mann, 1932; Shuman, 1950). These crucial nutrients can be distributed hundreds of kilometres inland from streams, even into upland forests (Reimchen, 2000). Estimates of sockeye salmon returns in Bristol Bay, Alaska, predict 20 million salmon during large years, producing over 54 million kilograms of biomass (Gende, Edwards, Willson, & Wipfli, 2002). Their role as agents of nutrient transfer between marine, aquatic and terrestrial systems means that the lives of thousands of individual organisms depend on healthy salmon runs and the resources they provide (Willson, Gende, & Marstron, 1998; Cederholm, Kunze, Murota, & Sibatani, 1999).

Additionally, the chinook, coho, and sockeye salmon are of particular importance to commercial and recreational industries (Hughes, 2013). Commercial harvest from the Cook Inlet alone brought in more than $10 million U.S. dollars in 2010 (Shields & Dupuis, 2012). Recreational fishing provides additional revenue, having generated $29 million dollars in 1986, and are estimated to have increased by 15% to 25% between 1986 and 2003, a trend that is expected to continue (Sweet, Ivey, & Rutz, 2003). However, both commercial and recreational revenues are dependent on seasonal spawning returns, which are influenced by the availability of suitable spawning habitat. Within the MSB, the availability of high quality, suitable spawning habitat is threatened by rapid urbanization and extraction of natural resources, both of which have the potential to seriously impact local salmon freshwater life stages (Stromberg & Scholz, 2011; Alderman, Lin, Farrell, Kennedy, & Gillis, 2016).

Anchorage, Alaska’s largest city, resides at the confluence of the MSB drainage and the Cook Inlet to the Pacific Ocean. The proximity of this metropolitan region to the salmon-bearing tributaries of the MSB has increased the anthropogenic impairment of salmon habitat. As of
2000, 42% of all Alaskans lived within the Anchorage municipal boundaries (Municipality of Anchorage, 2001). Anchorage accounted for almost half of the state’s population growth during the 1990s, and the area’s rate of growth is faster than the majority of metropolitan areas in the United States (Municipality of Anchorage, 2001). Between 2001 and 2009, this trend continued; 41.3% of the state’s growth occurred in Anchorage, and 34.1% of the state’s growth occurred in the MSB (Keith, Erben, & Dapcevich, 2010). Together, the growth of Anchorage and the MSB accounted for 74.4% of the state’s growth between 2001 and 2009. Development in the MSB has been ‘out not up’, with residential buildings sprawling beyond established communities, as many residents desire to make their homes adjacent to streams and lakes. An estimated 31% of MSB residents commute to Anchorage. Due to the rural demand for housing, agricultural land is being converted for residential development and retail (Mat-Su Salmon Partnership, 2013).

With increasing urbanization in the MSB, several anthropogenic impacts on the environment have threatened salmon spawning habitat. Loss of wetlands and riparian habitat, reductions in water quality and quantity, all terrain vehicle (ATV) use within stream channels, and culvert installation, have all concerned the Alaska Department of Fish and Game (ADF&G) as human caused impacts on salmon habitat (Hughes, 2013). Not only are urban land use changes responsible for habitat impairment, but also oil, gas, and mining operations jeopardize freshwater salmon habitat.

Rich, high quality mineral deposits remain an untapped resource for the MSB, with the greatest mining potential being rich coal deposits. Recent estimates from the Usibelli Corporation predict an annual yield of 500,000-700,000 tons (approximately 453,000-635,000 metric tonnes) in coal.
production spanning twelve years (Mativa & Hanson, 2008). As of September 2016, Alaska Department of Natural Resources Division of Mining renewed Usibelli’s mineral lease to this coal deposit (Hollander, 2014), and two additional mine proposals target the same coal deposit. As large mining operations remove mass from a drainage, groundwater flow paths, water quality, sediment transport, and fish access to habitat all become altered (Mat-Su Salmon Partnership, 2013). In addition to mining coal, companies are pursuing coal-bed methane extraction. A 2007 pilot project by Fowler Oil and Gas Corporation started tapping the existing reserves (Mativa & Hanson, 2008). Installation of well pads, roads and pipelines can lead to habitat fragmentation and sedimentation. Furthermore, accidental spills present unpredictable environmental risks associated with extractive resource development (Brittingham, Maloney, Farag, Harper, & Bowen, 2014). The presence of extractive industries in the landscape make necessary to quantify how different attitudes towards risk affect the chances of conservation success. Specifically, conservationists need to address whether effective conservation of salmon habitat can take place by just avoiding areas where extractive industries are present.

To maximize conservation efforts in landscapes facing anthropogenic development, systematic landscape planning software can be applied to provide cost effective, prioritized conservation solutions to optimize conservation investments. Systematic landscape planning software originally focused on conservation in terrestrial and marine ecosystems, however applications to lotic ecosystems require additional modifications. By applying existing terrestrial and marine procedures, protected areas may be clustered across catchment boundaries, not defined by stream networks. Failing to include the flowing nature of lotic ecosystems means that the solutions generated do not account for the connective habitat requirement of some riverine species,
especially species with large ranges (Fausch, Torgersen, Baxter, & Li, 2002). Fortunately, several authors have clarified topological rules to better represent the connectivity between upstream and downstream habitats, increasing systematic conservation planning applications to lotic ecosystems (Hermoso, Linke, Prenda, & Possingham, 2011; Esselman & Allan, 2011; Linke et al., 2012).

Using systematic conservation planning techniques, a series of scenarios were developed to determine management priorities for salmon spawning habitat conservation, including how spawning habitat is impacted by urbanization, oil and gas, and coal development. Four distinct scenarios were developed to test how different risk sources influence spawning habitat conservation priorities:

- Ignoring all anthropogenic risks to habitat, both urbanization and fossil fuel extraction (‘Risk ignored’)
- Accounting for risk associated with urbanization, avoiding all areas with fossil fuel extraction and deposits (‘Urbanization accounted, all extraction avoided’)
- Accounting for risk associated with urbanization, avoiding all areas with coal extraction and deposits (‘Urbanization accounted, coal areas avoided’)
- Accounting for risks associated with both urbanization and fossil fuel extraction, all areas are however available for conservation (‘All risks accounted’)

Naidoo et al. (2006) established that incorporating economics into conservation plans yield greater biological gains over plans ignoring costs. Therefore, land use data were used to calculate opportunity costs (in terms of lost potential revenue) of designating areas for conservation. These
land costs were then combined with data on spawning habitat locations to ultimately identify areas that represent conservation priorities under each scenario.

2. Methods:

Conservation Planning Overview

Marxan with probability optimization software was used in conjunction with environmental risk surface (ERS) models to identify priority salmon spawning habitat. (Fig. 1). Marxan software offers conservation planners decision support by optimizing which areas should be set aside for conservation to achieve a desired conservation goal (Possingham, Wilson, Andelman, & Vynne, 2006; Moilanen, Wilson, & Possingham, 2009). Within a Marxan analysis, the landscape is initially divided into ‘planning units’, areas at which management actions are undertaken. Marxan then selects a number of planning units from the total available and calculates whether pre-determined conservation targets (i.e. 30% of a species’ distribution) have been met. Using a simulated annealing optimization algorithm, Marxan then changes some of the selected planning units and calculates whether the change represents an improvement either in terms of conservation targets met or cost. If the newly selected planning units represent an improvement, the process is repeated. If the new planning units do not represent an improvement, the algorithm returns to the previous set of planning units and the process is repeated. Through this iterative process, Marxan can arrive at a set of planning units that achieve all conservation targets at a low cost. Additionally, by implementing Marxan with probability, risks are added as an extra data layer within the analysis, and can be independently minimized, similar to how costs are minimized. By including risks into the Marxan selection process, the risk of failure can be
included into how Marxan identifies an output reserve network (Tulloch et al., 2013), making the eventual solution more resilient to potential detrimental processes (Hammill et al., 2016). In this study, each Marxan scenario consists of 100 repeat runs, with 1,000,000 iterations being undertaking in each run, where solutions offer 95% certainty. While recent advances in freshwater systematic conservation planning present methods for implementing multiple zones, multiple actions, and multiple action and threat combinations (Moilanen, Leathwick, & Quinn, 2011; Cattarino, Hermoso, Carwardine, Kennard, & Linke, 2015; Hermoso, Cattarino, Kennard, Watts, & Linke, 2015; Cattarino et al., 2016), these methods do not include protocols for incorporating the risk of conservation actions failing. In the study presented here, understanding and simulating the risk of conservation actions failing was critical to comparing how scenarios that accounted for risk perform compared to scenarios that ignored risk.

Study Area

The MSB was subdivided into tributary sized basins, each of which represented a single planning unit (n=519) within the Marxan analysis. Tributary basins were derived from hydrologic unit code (HUC 12) basins. The HUC system uses a hierarchical system for assigning catchment sizes. HUC 12 basins capture tributary systems, which can be grouped into larger HUC 8 subbasins, representing medium-sized river basins. The system scales up to HUC 2 regions, outlining large river drainages (EnviroAtlas, 2017). Distributions of Pacific salmon spawning habitat were obtained through the Alaska Department of Natural Resources and spatially correlated with HUC 12 watersheds (Alaska Department of Natural Resources, 2017). The financial costs associated with setting aside a planning unit for conservation were quantified from available land cover data. Land costs associated with urban, agricultural, and undeveloped
areas were derived from existing parcel costs, as cost per acre, then correlated to corresponding land cover types in the United States Geological Survey Land Cover dataset to determine the spatial distribution of costs (Fig. 2a). Anthropogenic risks to salmon habitat (Fig. 2b) were assessed using an ERS model. ERS models synthesize relevant land uses based on impact intensity, and impact distance to clarify the extent of human caused impacts on the environment (McPherson et al., 2008). This process integrates into Marxan to minimize risks when identifying priority conservation areas (Lessman, Muñoz, & Bonaccorso, 2014; Evans, Schill, & Raber, 2015). Risk sources were compiled from urbanized landscape features included residential development, roads, and the threat posed by agriculture. Where applicable, these risks were combined with site-specific risks from mining and oil and gas development (Fig. 2c). Schill and Raber (2008) suggest incorporating risk accumulation in stream networks by applying ERS models to a flow accumulation simulation, as stressors to freshwater ecosystems may originate in distant upstream sources (Fig 1.b) (Lake, 1980; Skelton, Cambray, Lombard, & Benn, 1995; Moyle & Randall, 1998; Pringle, Scatena, Paaby-Hansen, & Nunez-Ferrera, 2000). This process specifies the path that risk flows across the landscape. Esselman and Allan (2011) successfully implemented this modification to address risks to streams in Mesoamerican streams, representing an early application of risk assessment within freshwater systematic conservation planning, offering guidance for this study. Following this procedure, a risk accumulation layer was developed from the ERS model to be input into Marxan with Probability.

Marxan with probability Setup

Protected area connectivity may be customized within the Marxan software. In the most basic form of Marxan, connectivity is customized using a boundary length modifier (BLM), which
regulates the compactness of the resulting conservation network based on the perimeter of selected priority areas (Ball, Possingham, & Watts, 2009; Fischer et al., 2010). Adjusting BLM values influences the fragmentation or continuity of the output conservation network, where lower BLM scores produce less connected output networks and vice versa. Despite the customization of these variables, applications of systematic conservation planning across varying ecosystems presents issues. Originally designed for terrestrial and marine conservation, applications of systematic conservation planning to lotic freshwater systems have been plagued by several shortcomings (Abell, Allan, & Lehner, 2007; Ball et al., 2009). First, calculations of boundary lengths based on an entire study area do not account for hierarchical stream orders within a river basin. By applying existing terrestrial and marine procedures, protected areas may be clustered across catchment boundaries, not defined by stream networks. Several authors have proposed modifications for integrating the linear nature of freshwater connectivity into existing systematic conservation planning software (Hermoso et al., 2011; Esselman & Allan, 2011; Linke et al., 2012). Of these, Esselman and Allan subdivided natural catchment boundaries into planning units and then calculated neighboring boundary lengths at a larger basin size (2011). By identifying boundaries within subbasins, then reconnecting subbasins within a study area, BLM values identify neighboring planning units within each subbasin for all subbasins across the landscape of interest (Esselman & Allan, 2011). However, this reconnection of small basins within a larger basin still does not distinguish between upstream and downstream connections. Hermoso et al., (2011) first established the rule for distinguishing connectivity. Next, Linke et al. (2012) improved to the field by clarifying more strict topological rules, utilizing the Pfafstetter stream classification scheme to refine stream network relationships and minimize distances between protected areas. Pfafstetter topological rules for stream networks were compiled from
the World Wildlife Fund’s HydroBASIN database and joined to the study area’s HUC 12 catchments (Lehner & Grill, 2013). The Pfafstetter rules for stream network connectivity were applied to this study for assessing connectivity in defining management priority areas, allowing for the crucial distinction between upstream and downstream connectivity.

Scenario Design

After establishing Marxan inputs and connectivity rules for the analysis, BLM modifiers were tested through a sensitivity analysis to determine the most cost effective and connective matrix of management priorities. Before splitting the analysis into four scenarios the best BLM value for the connectivity rules was determined. At a BLM value of one, the Pfafstetter settings had more connections and a cheaper cost than when no connectivity settings were applied. Therefore, a BLM value of one was held constant for testing all scenarios. For each of the four scenarios, a range of conservation targets were tested for each scenario, ranging from 10% to 40% of each species’ current distribution, at 10% increments. Ultimately, a conservation target of 30% was selected for the final comparison following Betts and Villard (2009), and due to increasingly missed targets above the 30% threshold. In the Risk ignorant scenario, Marxan was set to ignore anthropogenic risks to salmon spawning habitat and had no aversion to identifying priority conservation areas where oil, gas, and coal deposits were abundant, meaning that conservation decisions were based solely on cost and species distributions. In the Urbanization accounted, all extraction avoided scenario, Marxan was set to account for the anthropogenic risks associated with urbanization identified through the ERS model, while completely avoiding areas rich in oil, gas and coal deposits. Similar to the extraction-avoiding scenario, the Urbanization accounted, coal areas avoided scenario, Marxan was set to account for the anthropogenic risks associated
with urbanization, while completely avoiding areas rich in coal deposits. In the All risks accounted scenario, Marxan was set to account for all anthropogenic risks identified through the ERS, including urbanization and fossil fuel extraction. In this scenario, areas where oil, gas, and coal deposits were abundant were available for inclusion in a conservation network, but the risks to salmon habitat associated with these areas were accounted for in the selection process. Each scenario therefore represents a different attitude towards the different risks present on the landscape, and as a result, threats to the conservation success of each scenario are dependent on how threats manifest.

To compare the conservation success and resiliency of each scenario, risk was simulated for each scenario’s best solution from Marxan to determine how each scenario would likely perform in the face of conservation threats. Risk was simulated across the landscape-level conservation solutions generated from each of the four scenarios using Monte Carlo numerical simulations (Hammill et al., 2016). Risk was simulated over 1000 iterations, where for each iteration a random number was assigned to each planning unit. If the random number was less than the existing risk assigned to that unit (as defined by the ERS model) the planning unit was deemed ‘lost’ and removed from the scenario’s conservation solution. As a result, the removal of planning units subtracts from the total area protected over the landscape, potentially meaning insufficient planning units remain ‘not lost’ to meet the conservation target. By comparing the ratio of conservation targets met after risk simulation to the cost of implementing the conservation solution, a return on investment was calculated for the landscape solutions generated from each of the four scenarios.
3. Results

Each scenario addressed conservation risks differently, demonstrating the importance of attitude to risk on conservation success. The Risk ignored scenario identified management priorities without accounting for threats from anthropogenic activity or avoiding areas rich in extractive resources (Fig. 3a). In the absence of landscape level risk, the Risk ignored scenario would meet the defined 30% conservation targets for all five Pacific salmon species, at an estimated cost of $45,000 (Fig. 4a). However, when the predicted impact of anthropogenic activities was simulated, the predicted loss of planning units suggests that the solution would only protect 1.67 [SD, 0.08] species (Fig. 4b) due to the number of planning units predicted to be impacted by human encroachment, or extractive resource development. The Risk ignored scenario would therefore yield a return on investment of 0.39 [SD, 0.02] targets met per $10K spent (Fig. 4c).

Under an Urbanization accounted, all extraction avoided scenario (Fig. 3b), where risks associated with urbanization are accounted for in the Marxan analysis but areas with fossil fuels are unavailable for selection, 0 [SD 0.0] targets would be met (Fig. 4a), at an estimated cost of $98,000 (Fig. 4b). The Urbanization accounted, all extraction avoided scenario would therefore yield a return on investment of 0 [SD, 0.0] targets met per $10K spent (Fig. 4c). Under an Urbanization accounted, coal areas avoided scenario (Fig. 3c), where risks associated with urbanization are accounted for in the Marxan analysis but areas with rich in coal resources are unavailable for selection, 0.97 [SD, 0.02] targets would be met (Fig. 4a), at an estimated cost of $113,000 (Fig. 4b). The Urbanization accounted, coal areas avoided scenario would therefore yield a return on investment of 0.085 [SD, 0.002] targets met per $10K spent (Fig. 4c).

Following a simulation of landscape level risks, the All risks accounted scenario (Fig. 3d) would
meet an average of 4.73 [SD, 0.05] conservation targets (Fig. 4a) at an estimated cost of $58,000 (Fig. 4b). The All risks accounted scenario is therefore predicted to yield the greatest return on investment of 0.81 [SD, 0.009] targets met per $10K spent (Fig. 4c). Additionally, risk simulations were conducted for each scenario at 10%, 20% and 40% targets. At a 10% target all scenarios performed best, reaching the greatest return on investments. However, as targets were increased, the ability for each scenario to meet the targets decreased, and costs increased. The All risks accounted scenario was the only scenario able to maintain the number of targets met after risk was simulated onto the solution. However, increases in cost as targets increased, lead to overall decreases in return on investment, even for the All risks accounted scenario (Fig. 5).

Once targets reached 40%, both the Coal areas avoided, and All extraction avoided scenarios missed targets for all species and return on investments dropped to 0.

4. Discussion

With increasing anthropogenic stresses being placed on formally pristine habitats, it is critical to investigate how risk of human encroachment should be incorporated into conservation planning (Goudie & Viles, 2003). Results from this study demonstrate that simply choosing to ignore anthropogenic risk, and base conservation decisions solely on costs and species’ distributions represents a poor attitude towards risk as losses incurred prevent conservation targets being met. In addition, simply choosing to avoid locations with containing potentially catastrophic threats means that large portions on the landscape will be excluded, making conservation targets impossible to meet. This was seen as targets increased from 30% to 40%, the Coal areas avoided and All extraction avoided scenarios, all targets were missed. It is proposed that when making conservation decisions, the best attitude towards risk appears to be a willingness to accept risk
(i.e. do not simply avoid potentially risky areas) but incorporate this risk into conservation decisions (Hammill et al., 2016).

Under a Risk ignorant scenario, landscape decisions were based solely on cost and biodiversity data alone. While the solution generated through the Risk ignorant scenario at a target of 30% had the lowest up front cost, the number of conservation targets met following a risk simulation (1.67) was lower than the All risks accounted scenario (4.73) that accommodated for landscape risk. This low number of targets met is due to selected planning units being deemed ‘lost’ so that insufficient areas remain to meet conservation targets. The low number of targets met mean that a Risk ignorant strategy had a lower overall return on investment (0.39 targets met per $10K spent) than the All risks accounted scenario (0.81 targets met per $10K spent).

Under the Urbanization accounted, all extraction avoided scenario, and the Urbanization accounted, coal areas avoided scenario, large numbers of available planning units were locked out from possible solutions. Simply avoiding areas with fossil fuel development excludes a large portion of the landscape, making it impossible to meet conservation targets. In addition, although the solutions generated under the extraction avoided, and coal areas avoided scenarios did not meet all targets even before risk was simulated, both incurred higher upfront cost than the remaining scenarios. These high costs may be because the exclusion of large areas substantially reduces the options available, forcing the software to include expensive, sub-optimal planning units in the solution in an attempt to meet at least some conservation targets. These high costs also mean that the return on investment predicted to be obtained through the extraction avoided, and coal areas avoided scenarios were the lowest.
Finally, under the All risks accounted scenario landscape decisions incorporated cost, biodiversity data, while minimizing risks. Unlike the scenarios that merely excluded areas with extractive resources present, the All risks accounted scenario accepted risk associated with extractive regions and included that risk into the optimization process. Therefore, the resulting solution maximized return on investment as well as minimizing landscape risk, providing ‘risk proofing’ for the scenario. Due to the initial ‘risk proofing’ of the All risks accounted scenario, the Monte Carlo risk simulation affected this scenario less than the other three scenarios. The risk simulation for the All risks accounted scenario removed fewer planning units from desired targets, compared to the other three scenarios. Though the All risks accounted scenario incurred a greater upfront cost than the Risk ignored scenario, the All risks accounted scenario met more targets and yielded the greatest return on investment than the other three scenarios tested. Though the All risks accounted scenario was 29.8% more costly than the Risk ignorant scenario at the 30% target, the return on investment for the All risks accounted scenario was twice as large. By including potential anthropogenic risk factors, the All risks accounted scenario identifies priority areas of increased resiliency compared to priority areas identified when risks are ignored. As targets were increased from 10% to 40%, the All risks accounted scenario was the only scenario able to maintain the number of met targets following simulated risk across the study area. The high number of missed targets under both the Urbanization accounted, all extraction avoided scenario and the Urbanization accounted, coal areas avoided scenario suggests that coordinating effective freshwater salmon conservation in the MSB cannot be achieved by attempting to completely avoid areas rich in extractive resources. Managers may be pre-disposed to adopting risk averse attitudes towards conservation due to fear of failure.
(Maguire & Albright, 2005; Lennox & Armworth, 2011; Tulloch et al., 2015). However, results indicated that greater returns are obtained when managers accept certain risks into their salmon conservation strategies, and acknowledge that future energy extraction will influence freshwater salmon conservation.

Future efforts to improve the resiliency of salmon conservation in the MSB would be improved through increased data resolution. This study does not clarify how conservation priorities would change from fluctuations to yearly spawning returns. Spawning data provided by Alaska Department of Natural Resources clarified the spatial extent of spawning habitat, but did not clarify the density of redds in spawning areas. Nonetheless, in years with low spawning returns, fish use the same habitat as spawners from greater returning years, but in lower frequency. Therefore, the spatial priorities identified within this study apply for both high and low spawning return years, however the absolute magnitude of spawners is not included. Oceanic conditions have great influence on salmon productivity and mortality; driven by the Pacific Decadal Oscillation (Hare & Francis, 1995; Beamish et al., 2010). This paper does not suggest that the pelagic life stages of Pacific salmon are less vital for salmon conservation, but instead focused on the novel threats to freshwater salmon habitat from rapidly increasing human activity.

Management Recommendation:

Commercial and sport fishing represent multi-million dollar industries for Alaska, and the MSB is no exception. Fishing industries are bound by the success of seasonal salmon spawning runs and the health of freshwater salmon habitat. Meanwhile, human activities threaten critical
freshwater salmon habitat. By providing economically rooted conservation prioritization, this study intends to provide local managers and conservation groups with useful information to identify conservation opportunities in local river basins conflicted by land uses. The Urbanization risk included scenario suggests that risk adverse management techniques are impractical. The All risks accounted scenario highlights how including anthropogenic risks identify management priorities. The cost increase associated with accounting for All Risk (estimated $13,000.00) suggests that including risk into management decisions is achievable at a known price. Local non-profit Great Land Trust has been independently developing salmon conservation priorities for the MSB using different prioritization methods. The authors of this paper hope to share their results with both Great Land Trust and other local agencies, to work towards integrating conservation strategies for MSB salmon.

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References:


Figure legends

**Fig. 1.** Flow chart of the methods implemented in this study. Four distinct scenarios were tested, 1) Risk ignored; 2) Urbanization accounted, all extraction avoided; 3) Urbanization accounted, coal areas avoided; 4) All risks accounted.

**Fig 2.** Spatial distributions of data incorporated into Marxan analysis. (a) Land costs based on available land cover data, land costs are calculated per hectare in US dollars. (b) Distribution of environmental risks derived from ERS model. Inset describes how risk accumulation flows through stream network. (c) Fossil fuel resources within the Matanuska-Susitna Basin.

**Fig 3.** Planning units selected for the best solution under each Marxan scenario (a) Risk ignored scenario, (b) Urbanization accounted, all extraction avoided, (c), Urbanization accounted, coal areas avoided, (d) All risks accounted.

**Fig. 4.** Results summary for the four different risk scenarios following simulation of the impacts of environmental risk, (a) Number of conservation targets met, (b) Cost of best solution, (c) Return on investment.

**Fig. 5.** Results summary for the four different risk scenarios following simulation of the impacts of environmental risk tested at targets from 10% to 40%, (a) Number of conservation targets met, (b) Cost of best solution, (c) Return on investment.
Figure 1

Inputs
- HUC 12 catchments
- Salmon Spawning Habitat
- Land Cover Land Costs
- Roads Pipelines Agriculture Urban Areas

Pre-processing
- Pfafstetter rules
- Cost/Unit Area

ERS Model
- Risk Flow Accumulation

Setup
- Planning Units
- Biological Features
- Costs
- Risks

Marxan: Sensitivity Analysis

Analysis
- Marxan: Scenarios

Target values

Post-processing
- Monte Carlo Simulation: Scenarios

BLM values

Results
Figure 3

(a)  
(b)  
(c)  
(d)  

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