



Assessment method for epithermal gold deposits in northeast Washington State using weights-of-evidence GIS modeling

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Digital data and files

Text and data from this report can be found on the internet at the USGS website:
<http://geopubs.wr.usgs.gov/open-file/of01-501>

Text and other files accompanying this report are listed below:

[txt -- ascii text document, PDF -- Adobe Acrobat Portable Document Format, JPEG -- raster image compression formats]

Name	Folder	Type	Description	Size, MB
1_README.txt	Epithermal_OFR	ascii	readme	0.07
OF01-501.jpg	Epithermal_OFR/Report	PDF	Report text and tables	0.91
fig1, fig4ab, fig6, fig9, fig10, fig12	jpg	JPEG	Figures 1, 4, 6, 9, 10, 12	2.22
fig1, fig4ab, fig6, fig9, fig10, fig12	pdf	PDF	Figures 1, 4, 6, 9, 10, 12 (600 dpi)	3.08

ABSTRACT

The weights-of-evidence analysis, a quantitative mineral resource mapping tool, is used to delineate favorable areas for epithermal gold deposits and to predict future exploration activity of the mineral industry for similar deposits in a four-county area (222 x 277 km), including the Okanogan and Colville National Forests of northeastern Washington. Modeling is applied in six steps: (1) building a spatial digital database, (2) extracting predictive evidence for a particular deposit, based on an exploration model, (3) calculating relative weights for each predictive map, (4) combining the geologic evidence maps to predict the location of undiscovered mineral resources and (5) measuring the intensity of recent exploration activity by use of mining claims on federal lands, and (6) combining mineral resource and exploration activity into an assessment model of future mining activity.

The analysis is accomplished on a personal computer using ArcView GIS platform with Spatial Analyst and Weights-of-Evidence software. In accord with the *descriptive model* for epithermal gold deposits, digital geologic *evidential themes* assembled include lithologic map units, thrust faults, normal faults, and igneous dikes. Similarly, geochemical evidential themes include placer gold deposits and gold and silver analyses from stream sediment (silt) samples from National Forest lands. Fifty mines, prospects, or occurrences of epithermal gold deposits, the *training set*, define the appropriate areally-associated terrane. The areal (or spatial) correlation of each evidential theme with the training set yield predictor theme maps for lithology, placer sites and normal faults. The weights-of-evidence analysis disqualified the thrust fault, dike, and gold and silver silt analyses evidential themes because they lacked spatial correlation with the training set. The decision to accept or reject evidential themes as predictors is assisted by considering probabilistic data consisting of weights and contrast values calculated for themes according to areal correlation with the training sites. Predictor themes having acceptable weights and contrast values are combined into a preliminary *model* to predict the locations of undiscovered epithermal gold deposits. This model facilitates ranking of tracts as non-permissive, permissive or favorable categories based on exclusionary, passive, and active criteria through evaluation of probabilistic data provided by interaction of predictor themes. The method is very similar to the visual inspection method of drawing conclusions from anomalies on a manually overlain system of maps. This method serves as a model for future mineral assessment procedures because of its objective nature.

To develop a model to predict future exploration activity, the locations of lode mining claims were summarized for 1980, 1985, 1990, and 1996. Land parcels containing *historic* claims were identified either as those with mining claims present in 1980 or valid claims present in 1985. *Current* claim parcels were identified as those containing valid lode claims in either 1990 or 1996. A *consistent* parcel contains both historic and current claims.

The epithermal gold and mining claim activity models were combined into an assessment (or mineral resource-activity) model to assist in land use decisions by providing a prediction of mineral exploration activity on federal land in the next decade. Ranks in the assessment model are: (1) no activity, (2) low activity, (3) low to moderate activity, (4) moderate activity and (5) high activity.

INTRODUCTION

1. Objectives and purpose

This report presents the results of an assessment of the relative favorability and anticipated future minerals development activity in northeastern Washington for gold-silver-bearing epithermal mineral deposits. The two-part assessment is accomplished using weights-of-evidence in a geographic information system (GIS). The weights-of-evidence analysis is a data driven GIS tool adapted from the medical diagnostics field to mineral resource assessment (Kemp and others, 1999).

The first component of the assessment is based on the physical evidence (geology, geochemistry, geophysics and so on) that indicates mineralizing processes may have occurred in an area. In this case, the epithermal gold deposit model incorporates geological and geochemical data in assessing the probability of occurrence of undiscovered gold and silver deposits. A procedure is introduced in the construction of this model using the probability results from the weights-of-evidence analysis as a decision-making tool. This tool is used to establish the relative favorability of tracts for the occurrence of mineral deposits using the terms non-permissive, permissive, and favorable tracts. The exploration activity model incorporates mineral-industry exploration activity using past and current mining claims data, industry reports, publications, and other public-domain resources as a predictor of the minerals exploration and development activity which may be expected in a given area. The significance of the two-part approach to assessment is that it combines the results of both the undiscovered mineral resource and mining claim activity.

This study was performed to assist the United States Forest Service in planning for land uses. The study area includes much of the Okanogan and Colville National Forests. Six related geologic reports provided supporting digital data for this analysis can be downloaded from the USGS website. The first report presents the geologic raster data (Boleneus and Causey, 2000)--lithology, faults, folds, and igneous dikes themes--used extensively in the weights-of-evidence analysis. The report documents the study area's seamless digital geologic map constructed from six, 100:000-scale geologic maps and includes ArcView shapefiles. Various geochemical themes were also prepared. Analytical data for gold, silver, lead, zinc, copper, molybdenum, tungsten, and uranium were digitally compiled for 3,927 rock and stream sediment (silt) samples collected by R.A Grant (unpublished data) from the Okanogan and Colville National Forests (Boleneus and Chase, 1999). Four other reports present geological and mineral activity databases used. The first includes the training set of 50 epithermal deposits and the 67 gold placer sites (Boleneus, 1999a). Hyndman and Campbell (1999) prepared the digital database of mining claims used for mineral industry exploration activity data. Two other reports document mineral industry exploration activity data in Washington examined during the analysis. These include mineral permits issued by the Colville, Kaniksu, Okanogan, and Wenatchee National Forests (Boleneus, 1999b) and a 13-year summary of mineral industry project activity in Washington (Boleneus and Derkey, 2000).

Also, the method serves as a model for future mineral assessments because of its objective nature and strict application of procedures of the weights-of-evidence analysis. Using the information and procedure applied here--epithermal deposit model, data, software, and methods--duplication of these results by others is assured.

Analysis was performed on a personal computer using ArcView v. 3.1³, Spatial Analyst v. 1.0 and Weights-of-Evidence (Kemp and others, 1999) software. The assessment method requires that all data be analyzed in digital form. At least 100 person-days were expended to prepare the data before the analysis. The authors accomplished the analysis in a committee setting during three days.

2. Location

The study encompasses a 222-km x 277-km area bounded by Idaho on the east and Canada on the north. It includes all of Pend Oreille County and most of Stevens, Ferry, and Okanogan counties (lat 48-49 N and long 117-120 E). The area occupies six, 1:100,000-scale U.S.G.S. quadrangles (Omak, Oroville, Colville, Chewelah, Republic, and Nespelem) in Washington State (fig. 1).

Exploration for gold has been concentrated around operating mines in Ferry and adjacent counties, including epithermal (or hot spring) gold deposits at Republic, Kettle, K-2 and Orient.

The report, in particular, addresses about one-half of the study area, or 23,600 km² that contains Eocene volcanogenic rocks that host the deposits. Epithermal gold deposits are, for the most part, restricted to those Eocene volcanogenic and sedimentary rocks contained within the Republic Graben, the largest of four fault-bounded extensional troughs (Full and Grantham, 1968; Fifarek, Devlin, and Tschauder, 1996). This graben occurs between two pre-Eocene-age core complexes, the Okanogan and Kettle gneiss domes, consisting of meta-igneous, meta-sedimentary, and meta-volcanic rocks whose protoliths range in age from Devonian to Cretaceous (Fox, 1994; Box, 1994).

3. Mineralization

Epithermal gold deposits in northeast Washington are described by several authors (Full and Grantham, 1968; Tschauder, 1989; Fifarek, Devlin, and Tschauder, 1996; and Rasmussen and Gelber (written communication, 2000, 2001). Deposits are formed in a near-surface environment by deposition of gold and silver, in quartz-pyrite-clay-carbonate (+/- calcite, marcasite, ankerite, illite, kaolinite, and alunite) veins in a hot-spring environment. Deposits occur within a graben-filling Eocene pyroclastic, fluvial, and lacustrine succession consisting of O'Brien Creek, Sanpoil Volcanics, and Klondike Mountain formations. These Eocene successions occur within each of at least four grabens developed during the later stages of emplacement of the Kettle and Okanogan gneiss domes.

The Knob Hill and Golden Promise deposits (collectively, the Hecla Mining Co. "Republic" operation) of the Republic district are the most notable and closed in 1995 after 100 years of activity. The Republic operation was located about 2 km northwest of the city of Republic, Washington. Golden Promise is located about 1.7 km southeast of Knob Hill. The Kettle and K-2 deposits had similar origins to the deposits at Republic and are located 25 km to the north-northeast. The Knob Hill mine was developed from a decline shaft but more recently, the K-2 and Golden Promise deposits were developed from decline or incline ramps and mined by cut-and-fill mine methods.

³ ArcView and Spatial Analyst are software products of Environmental Systems Research Institute (ESRI) Inc., Weights-of-evidence software is a product of Geological Survey of Canada

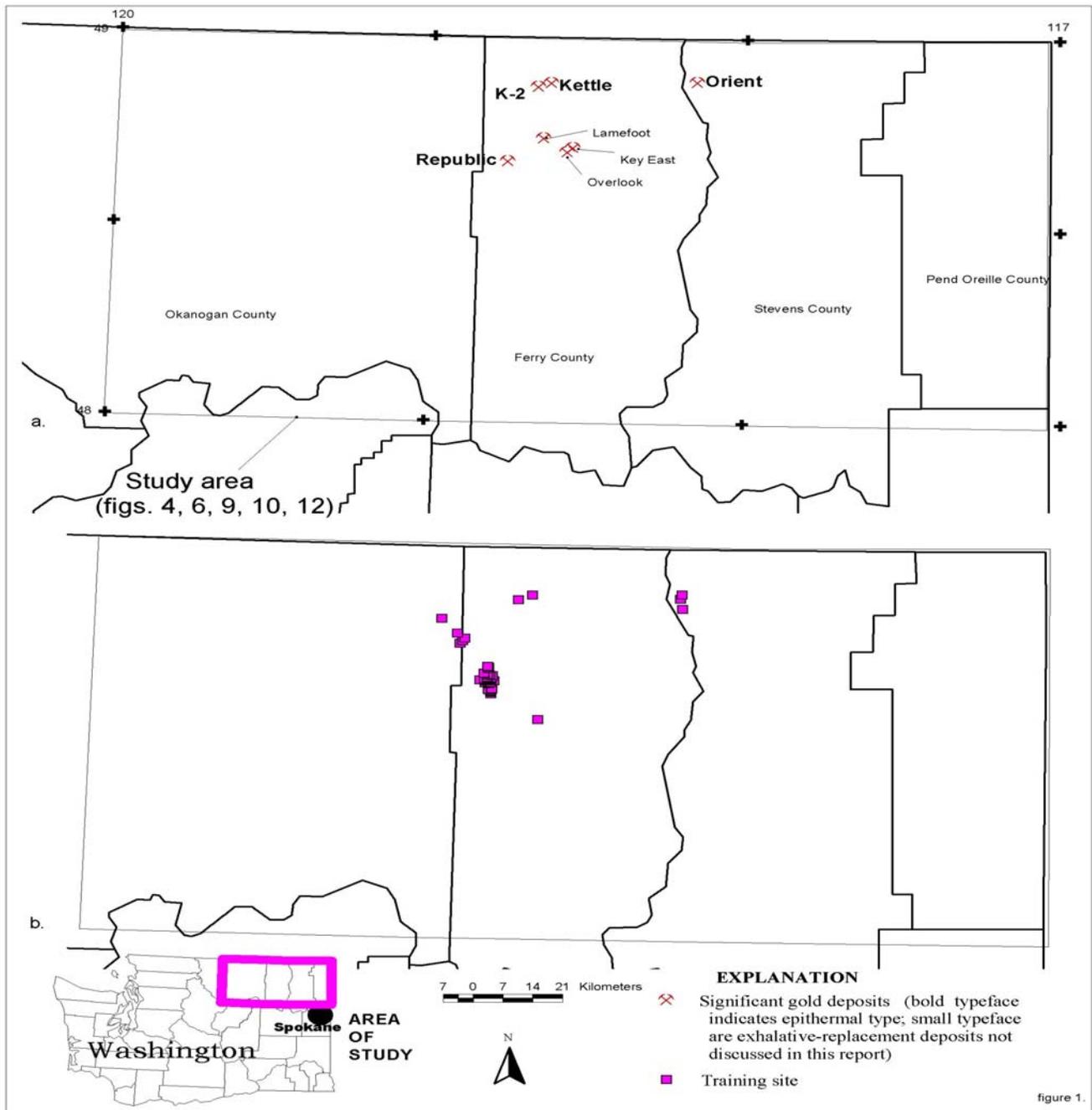


figure 1.

Figure 1. Location map showing training sites and significant epithermal gold deposits in northeast Washington.

(a) Significant epithermal gold deposits are Orient, Kettle, K-2, and Republic (Knob Hill). Other nearby gold deposits, the Lamefoot, Key East, and Overlook are exhalative-replacement gold deposits and are not considered in this report, but accounted for approximately 1 million troy ounces of gold production through 1997. (b) Fifty training sites are shown on the map. Training sites are locations of gold mines, prospects, or occurrences of epithermal origin. About one-half of the training sites have recorded gold production. See text for explanation of training sites.

The gold deposits occur within the Sanpoil Volcanics up to 5 km eastward of the NNE-trending Bacon Creek Fault. The Bacon Creek Fault forms the west flank of the Republic graben, and numerous lesser NNE and NW-trending en echelon faults host gold-silver-bearing veins formed as a result of dextral-shearing adjustments in the Republic graben. Deposits occur entirely within the Sanpoil Volcanics and terminate (sometimes at sinter deposits) near the unconformity with the overlying Klondike Mountain Formation. The Sanpoil Volcanics consists of andesite and dacite lavas, epiclastic and pyroclastic rocks, conglomerate, and hornblende andesite porphyry flows. Free gold, gold selenides, and silver sulfosalts occur in conduits, vents and eruption ejecta of a Golden Promise hot springs system (Fifarek, Devlin, and Tschauder, 1996). Golden Promise vent breccias contain near-vertical, anastomizing zones of disseminated and stringer ore flanked by low grade gold in sinter at higher level. Sinter gold zones occur near the top of the Sanpoil Volcanics while the deeper, higher grade veins are characterized by breccia and colloform quartz textures in faults and fault breccia zones.

At Republic, the gold-silver-bearing veins consist of colloform and brecciated quartz where quartz, calcite, and sulfosalt minerals resulted from hot spring activity. Clays include kaolinite and alunite. Coarse, free gold is common. Ag-Au ratio is 5:1. At K-2, the gold-silver-bearing, bladed latticework quartz and brecciated quartz predominates over colloform quartz textures. Also at K-2, little free gold exists and clays consist of illite and kaolinite. Sinter zones and sulfosalts are absent, ankerite predominates over calcite, and the Ag-Au ratio is 1:1 (Rasmussen and Gelber, written communication, 2000). The Kettle deposit is very similar to the Golden Promise deposit in both morphology and mineralogy (Rasmussen, written communication, 2000). These are identical to the well-known model for epithermal hot spring gold deposits presented by Buchanan (1981) for deposits of the southwest U.S.

Production

The principal mining district in the map area is the Republic Mining District. The first claims were staked here in 1896 and by 1901 two mills were operating. The first was a custom mill at the town of Republic, Washington and another, at the Mountain Lion mine, about 3 km northwest of Republic (Landes and others, 1902). About ten mines continuously operated in early history of the district. By 1912, the district had produced nearly \$5,000,000 in gold and silver. In 1936, owners of the Knob Hill Mine built a mill at the site of principal activity located about 2 km north of the town (Full and Grantham, 1968). Later known as the Republic mine, this property operated nearly continuously by Knob Hill Mines Inc., then by Day Mines Inc. and later by Hecla Mining Co. until closing in 1995. Historically, through 1997, the region has produced at least 3 million troy ounces (93 tonnes) gold and 17 million troy ounces (550 tonnes) silver from all deposits of epithermal origin (Fifarek, Devlin, and Tschauder, 1996). Echo Bay Mines, Inc. operated the Kettle mine for a short time before its reserve became exhausted and now operates the K-2 mine, the only active gold mine in the state.

4. Previous work

Use of the weights-of-evidence as an assessment method as applied to mineral deposits is a recent innovation. Bonham-Carter (1994) presents an in-depth explanation of the weights-of-evidence analysis and other GIS methods in geology and Wright and Bonham-Carter (1996)

applied the method to search for areas likely to contain massive sulfide deposits in greenstone terrenes of Manitoba, Canada. Weights-of-evidence analysis has been used to predict favorable areas for vein gold deposits in Canada (Bonham-Carter, Agterberg, and Wright, 1988), for Carlin and epithermal gold deposits in Nevada (Mihalasky, 1999), and epithermal deposits in the Great Basin of the western United States (Raines, 1999). Explanation of the software operations of weights-of-evidence method followed in this report is available in the user guide (Kemp and others, 1999) and Raines, Bonham-Carter, and Kemp (2000) present an overview of the method.

WEIGHTS OF EVIDENCE

1. Method

The process used in weights-of-evidence modeling essentially is a quantitative version of the inspection method of overlaying several different map themes to identify areas where mineralization may be present (fig. 2). In the inspection method, the larger the number and magnitude of appropriate overlapping anomalies in data maps such as geochemistry, geology, or others, the greater the qualitative indication that mineralization may be present. In weights-of-evidence modeling, the importance of theme layers in delineating areas with potential for deposits is determined mathematically by how it compares with the areal distribution of the training set. When several themes are combined, the areas with the greatest coincidence of weights produce the greatest probability of occurrence of undiscovered mineralization.

Briefly, weights-of-evidence analysis is a map-correlation and map-integration process that is applied by formulating mathematical odds for (and against) and combining this evidence in support a hypothesis. In this report evidence consists of various evidence themes (see evidential themes, in Glossary) of exploration data, and the hypothesis is “this location is favorable for occurrence of deposit type ‘X’”. The odds of this association between the training set and each exploration theme (data) are measured and expressed as “weights”, defined as the natural log of the odds. The Glossary gives an example of a weights and contrast calculation. Kemp and others (1999) and Bonham-Carter (1994) give a detailed treatment.

The weights-of-evidence analysis of epithermal gold deposits is applied here in two parts. First, the analysis assists in determining the limits for favorable, permissive, and non-permissive areas that may be used in exploring for undiscovered epithermal gold deposits. Second, a mining claim activity is derived and this, in turn, is combined with the epithermal gold model to create the assessment model. The assessment describes the probability of both the undiscovered mineral resource for gold and the mining claim activity. The preparation and analysis steps followed in this report are shown in a flow diagram (fig. 3) and briefly outlined below.

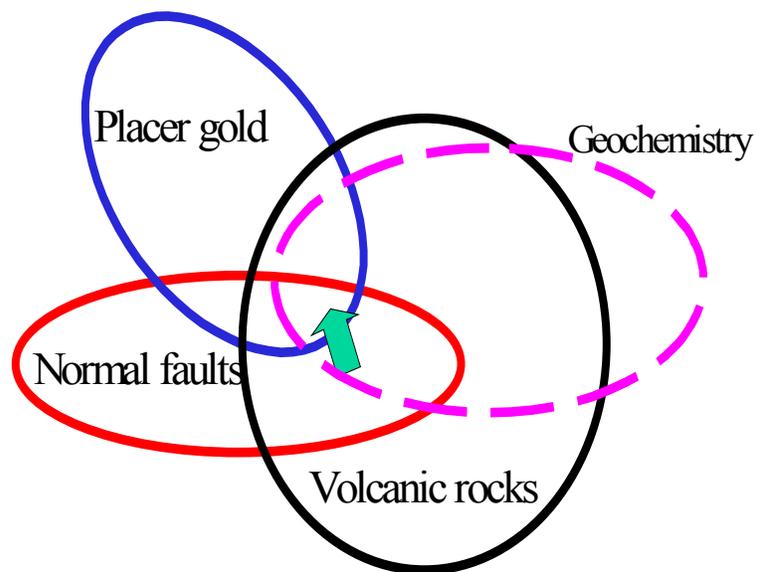
The procedure followed in applying a weights-of-evidence analysis in the first part of the two-part approach is carried out in five steps, as follows: (1) selection of a descriptive model; (2) selection of a training set; (3) selection of exploration (evidence) themes based on the descriptive model; (4) testing of the exploration themes to qualify them as viable (predictor) themes; and (5) consolidating the themes into a useful resource prediction model. The descriptive model used here is the mineral deposit model for epithermal (hot spring) gold deposits (Full and Grantham, 1968; Berger, 1986; Fifarek and others, 1996; M. Rasmussen and C. Gelber, written communication, 2000). These authors describe the geological setting and various characteristics of epithermal veins. Characteristics set out in the mineral deposit model are closely followed in selecting the training sites and evidence (or exploration) themes in the analysis. The training set consists of gold-silver mines (or non-producing prospects) of inferred epithermal origin. The

location and characteristics of these mines or prospects were assembled from published sources such as U.S. Geological Survey' MRDS (USGS, 2000) or MAS/MILS (USBM, 1995) databases. In step 3, the evidence themes were assembled from several geological, geochemical, and other databases that describe aspects of the mineral deposit model in the study area. Evidence themes assembled include a digital geological map (Boleneus and Causey, 2000) containing lithologic units, dikes, folds, and faults and digital geochemical themes consisting of placer gold mines (Boleneus, 1999a) and gold-silver stream silt analysis (Boleneus and Chase, 1999). The geological map was manually compiled at a scale of 1:100,000 by the Washington Department of Natural Resources (Wash. DNR) in the 1980's-1990's and converted into digital form. Geological data (lithology, faults, dikes) defined by polygons or lines were prepared as ArcView shapefiles from Arc/Info data supplied by Wash. DNR. Thrust faults and normal faults themes were from a larger set of faults data. Sources of the placer gold mine sites were the USGS MRDS and USBM MAS/MILS. Mine site, training site, and geochemical themes were prepared as point-type database files. Using gold and silver geochemistry (Boleneus and Chase, 1999) contour maps were interpolated from point-type data using Spatial Analyst software in conjunction with the ArcView GIS platform software. Shapefiles and point databases were converted to grid format and then a series of buffer bands (25m up to 1000m in width, as appropriate) were digitally constructed around the line, polygon, and point themes prior to the analysis. As many as 30 buffer bands were constructed with a radius ranging up to 30,000m outboard from the lines or polygons (as appropriate).

In step 4, the testing process consisted of digitally comparing the areal (or spatial) distribution of training set and evidence themes. Testing produced weights, contrast, and other statistical values calculated for each of the seven comparisons (lithologic units, normal faults, thrust faults, dikes, placer sites, stream silt gold, and stream silt silver). The weights (positive weight W^+ , negative weight, W^-) express the degree of spatial association between the training set and the evidence theme. The contrast value is merely the difference between the positive weight value and the negative weight value. The contrast value is the basis for accepting (or rejecting) the evidence themes as predictor themes. It evaluates the significance of each buffer band, that is, the level of area (spatial) correlation between exploration theme "X" and the training set. In step 5, the predictor themes are consolidated into an epithermal gold deposit model by addition of values at each grid cell. The statistical significance of the contrast can be assessed to decide that the observed contrast is not due to a random event. For this study a confidence of approximately 98% was required to define a significant contrast.

A procedure is introduced that uses the weights-of-evidence results as a decision-making tool to establish relative favorability (and unfavorability) of tracts for the occurrence of mineral deposits. In this report, favorable, permissive, and non-permissive tracts (areas) are defined by the degree of overlap of predictor themes in the epithermal gold deposit model. This concept of geological favorability of tracts is not a part of the weights-of-evidence procedure. The concept (Singer, 1993) expresses the probabilistic data of this analysis in a geological context (favorable, permissive, non-permissive) applicable to mineral deposits.

In the second part of the two-part approach, the objective is to combine the epithermal gold deposit model with the mining claim activity information to create the assessment model. The steps are: (1) obtain mining claim activity data for the years of 1980, 1985, 1990, and 1996, (2) create a matrix to combine the mining claim activity and epithermal gold model, and (3) derive an assessment model by applying this matrix. The second part uses ArcView and Spatial Analyst software but is not a weights-of-evidence analysis.



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Figure 2. Venn diagram showing overlay of geologic and geochemical themes used in weights-of-evidence analysis.

The probability that a deposit exists increases where the predictor themes areally overlap (green arrow).

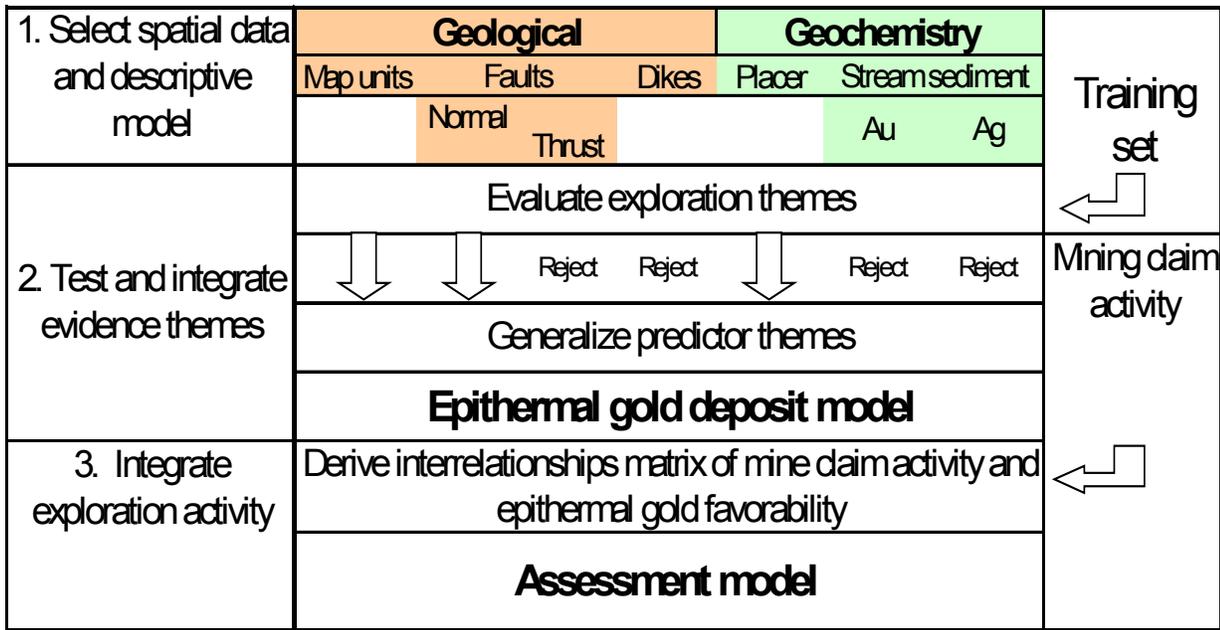


Figure 3. Flow diagram describing mineral assessment procedure for epithermal gold.

The first step in the analysis consists of assembling the evidence themes and training set for use in the analysis. The nature of this data depends on the descriptive model of the deposits sought. In step 2 of the analysis, each evidence theme is compared individually to the training set following the weights-of-evidence analysis procedures. Based on these results, themes are accepted or rejected based on calculated statistics (contrast, weights) that define probabilities. Acceptable themes of lithology, normal faults, and placer gold sites (predictors) are generalized based on an optimum buffer distance as indicated from the theme “peak” contrast values. The predictor themes are digitally added to form the model for undiscovered epithermal gold deposits. Using a probability approach, tracts are defined as non-permissive, permissive, and favorable for containing undiscovered epithermal gold deposits. In step 3, mining claim data are assembled and used to indicate levels of mineral industry exploration activity. The next step in the analysis involves determining the relationship between the level of mineral industry exploration activity and the non-permissive, permissive and favorable tracts. These themes are digitally added to form the assessment model of mineral industry activity for undiscovered epithermal gold deposits.

To prepare the mining claim activity theme, parcels containing mining claims were qualified as either containing historic (1980 or 1985), current (1990 or 1996), or consistent (current and historic). Historic, current, or consistent rankings occupy the x-axis on the resource model-activity model matrix. The y-axis is established by rank of favorable, permissive, or non-permissive from the epithermal gold deposit model. Classes within this matrix were assigned levels of mineral industry exploration activity consisting of no activity, low activity, moderate activity, and highest activity. The assessment model was created according to guidelines set out in the matrix.

2. Categorical or cumulative calculations?

Weights calculations for weights of evidence are carried out using one or both of two different procedures, the categorical or cumulative weights calculations. The available data dictates the procedure followed. The categorical method is used where data occurs in unrelated and mutually exclusive categories (nominal or classificatory scale of measurement of Siegel, 1956, p. 22). The categories Yes and No and lithologic units on a geologic map are examples. See Weights Calculation in App. II.

Unlike the unrelated data categories, where data between categories are related (ordinal, interval, or ratio data after Siegel, 1956, p. 22), the calculation may be carried out on a cumulative weights fashion. Data occurring in ranks or on a temperature scale are examples.

The basic procedure for calculating cumulative weights is to digitally establish a series of buffers around line, point, or polygon features of the exploration theme being tested and, during the hypothesis-testing step, to calculate weights and contrast values for each buffer band. Buffer bands are arbitrary and vary in width from one to 1000 or more meters. Contrast is the difference between the calculated positive and negative weights on the map for the exploration theme "X". The buffer band with the maximum contrast value is the optimal buffer of that theme. The buffer band containing the optimal buffer defines the limit (starting at the line, point, or polygon and moving outboard) for an evidence theme. Data that occur within the optimum buffer of a theme are termed to lie *inside the pattern* while all other data lie *outside the pattern*. Inside and outside describe binary theme generalization that uses the optimum buffer distance to define this limit (or boundary). Themes are referred to as "generalized" to the limit defined by the optimum buffer band. See glossary (app. II). The final step in the procedure is to add all predictor themes to create the final model.

3. Descriptive model for epithermal gold deposits

The descriptive ore deposit model guides the selection of: (1) the training sites and (2) the procedure for testing the evidence themes by the weights-of-evidence analysis. Definitive study of gold-silver epithermal deposits at Republic (Full and Grantham, 1968), Golden Promise (Fifarek and others, 1996), and K-2 (Rasmussen and Gelber, written communication, 2000) formed a descriptive model for selecting the training set and to guide the analysis.

The area has the following characteristics that are generally consistent with the USGS descriptive model for hot spring-type Au-Ag (Berger, 1986):

- Major features. The district lies at the center of a region in north-central Washington exhibiting extensive, generally dacitic, volcanic activity of mid-Eocene age. Evidence of this activity consists of a wide range of flows, pyroclastic rocks and water-laid tuffaceous

sediments. The northeast-trending Republic and Toroda Creek grabens, the major structural features preserve a large volume of these volcanic rocks. Movement along graben-bounding normal and low-angle, listric-normal faults appears to be coincident with cessation of volcanism, with uplift of the Kettle and Okanogan gneiss domes, and with gold mineralization. This suggests a genetic relationship between these events (Fox, 1994; Box, 1994). Gold deposits are formed in similar settings in southern California and southwest Arizona (Long, 1992).

- Alteration. Bleaching and alteration to clays is common; kaolinization of feldspars is related to widespread fracturing in the epithermal environment. Silicification is pronounced and adularia is present throughout the productive veins. Pyrite, chlorite, epidote and calcite are associated with hydrothermal alteration but the same minerals are not necessarily related to economic mineralization.
- Host rock. Veins favor hard or brittle rocks, such as intrusive porphyry or extrusive flow rocks. Veins, fault breccia, and hydrothermal breccia form along NW-trending, secondary, and sympathetic faults. Orientation of these and other faults can range from northeast to northwest. Banded, chalcedonic, and hydrothermal breccia veins that host the higher concentrations of gold and silver result from open-space filling in the boiling zone, and often are affected by post-mineral faulting. Low grade disseminated gold-silver values also occur in the porous, sericitized, pebbly zone of the basal lake bed sequence located stratigraphically above the veins.

TRAINING SITES

Training sites are a collection of mines, prospects or mineral occurrence sites having characteristics in common to those sought. Selection of the appropriate training sites within the limits of the study area is key to the analysis. Training sites were selected from the U.S. Geological Survey's Mineral Resource Database System (USGS, MRDS) (Boleneus, 1999a). Two other significant sites, Knob Hill Mine and K-2, were included from the US Bureau of Mines MAS/MILS database (USBM, 1995). These sites were sufficiently studied to infer confidently that they are related to epithermal gold mineralization. Informal rules for selecting a training sites dictate that chosen sites of epithermal gold mineralization are located both within the limits of the volcanic province of northeast Washington and within the study area. A set of training sites must be sufficiently large to obtain statistically significant results. Therefore simply selecting the significant deposits or only those exhibiting historic production did not provide enough geographically separated training sites. The number of sites selected is arbitrary. The fifty selected sites (App. I) that have common minimum characteristics could be related to either hot spring Au-Ag deposits (Berger, 1986), Creede epithermal deposits (Mosier and others, 1986), or Comstock epithermal vein deposits (Mosier, Singer, and Berger, 1986). See figure 1b. The area of each training site was assumed to be 1 km² for the purpose of the analysis. This provides a prior probability (p), the ratio of area of training sites to that of the area of study (50/23600), of 0.0021.

MRDS or MAS/MILS data indicated that 23 of the training sites are current or former producers, although this was not a requirement for selection. Five of these sites have produced significant quantities of ore (First Thought, Knob Hill, K-2, Kettle, Sheridan). Several sites (Ben Hur, Black Tail, Butte & Boston, El Caliph, Insurgent, Last Chance, Little Cove, Lone Pine, Morning Glory, Mountain Lion, Pearl, Princess Maude, Quilp, San Poil, Surprise, Republic, Tom

Thumb, and Trade Dollar) were small, but important producing sites that lay along the important 6.5-km-long Eureka fault trend (Full and Grantham, 1968) that extends northwest from the city of Republic. A large number of sites concentrated in a small area along the Eureka fault trend near Republic raised concern that a bias may be inadvertently introduced. Also mines adjacent to the Eureka fault are vein-type mines, but it is not known whether the training set is biased toward vein-type mines because information to make a judgment is not available from all sites. Randomized methods of site selection were not investigated. Golden Promise was not included in the set because an accurate location was unavailable at time of the analysis. Concern about its omission was dismissed because it lies within 300-500 m (or within the 1 km² site area) of three included sites, Black Tail, Surprise, and Last Chance.

ANALYSIS OF GEOLOGIC PATTERNS FOR EPITHERMAL MODEL

The objective of weights-of-evidence analysis is the *analysis of areal proximity* of the training set in comparison to other features (geological map features, geophysical features, alteration features, geochemical features and so) defined by points, lines or polygon areas. The results of three proximity analyses are described in this section. This analysis is not meant as an exhaustive treatment in this setting, or that it uses all available data, but merely demonstrates its strict application in a relatively well-studied location.

Based on the descriptive model for epithermal gold deposits in northeast Washington, several themes were assembled from various sources and tested as possible predictor themes (**table 1**). Themes tested and accepted as predictor themes, lithologic (volcanic) units, normal faults, gold placer sites, are in italic typeface; other themes were investigated but failed to meet the proximity criteria of the analyses. The acceptable theme data, called predictor themes, describe the lithologic, structural, and geochemical characteristics of the resource model as described below. A brief explanation is given in the next section of reason for failure of the igneous *dikes* (Eocene) theme to qualify as a predictor theme. The reason that the thrust faults and gold and silver silt geochemistry themes failed to qualify as predictor themes is beyond the report scope.

1. Proximity to lithologic units

Data preparation

Geologic map information in digital format was used to perform this analysis. Geologic theme data relied on 1:100,000-scale mapping completed by Washington Department of Natural Resources for Omak (Gulick and Korosec, 1990), Chewelah (Waggoner, 1992), Colville (Joseph, 1990a), Oroville (Stoffel, 1990a), Nespelem (Joseph, 1990b), and Republic (Stoffel, 1990b) quadrangles. Arc/Info coverages of these geologic quadrangles were converted into digital form at Wash. DNR (Schuster and Harris unpublished data) and provide to USGS in preliminary form. USGS divided these sub-coverages consisting of lithology, faults, folds, and dikes. The Wash. DNR coverages will eventually be published but may differ from those we used. Sub-coverages have spatial orientation links to a detailed data table containing topological and geologic data. Each of the four Arc/Info sub-coverages for the six quadrangles was combined into a seamless coverage corresponding to the limits of the study area. In the process, 670 lithologic units on the six quadrangle maps were generalized to 169 lithologic units in constructing the geologic map. All sub-coverages were converted to ArcView shape files and grid files (Boleneus and Causey,

2000). Because the many Quaternary units differed considerably between the 1:100,000-scale maps, they could not be correlated from map to map and were abandoned in construction of the geologic map. Instead, one Quaternary undivided (Q) unit was used to describe all Quaternary units.

Table 1. Themes investigated for epithermal gold deposit model.

[Themes in italic typeface were accepted as predictor themes. Other themes were tested but rejected as predictor themes.]

Theme	Purpose	Source of data
<u>Predictors:</u>		
<i>Lithologic units (i.e. Eocene volcanic rocks)</i>	<i>Specify extent of host rocks</i>	<i>Lithology coverage (Wash. DNR digital geology)</i>
<i>Normal (steep) faults</i>	<i>Indicate favored hosts for veins</i>	<i>Faults coverage (Wash. DNR digital geology)</i>
<i>Placer gold sites</i>	<i>Identify eroded gold deposits</i>	<i>MRDS (USGS), MAS/MILS (USBM)</i>
<u>Not predictors:</u>		
Dikes (Eocene)	Outline center of volcanic activity	Dikes coverage (Wash. DNR digital geology)
Low-angle, listric (“thrust”) faults	Denote deep crustal structures possibly related to detachment faults	Faults coverage (Wash. DNR digital geology)
Gold and silver assays from stream sediment samples	Identify eroded gold deposits	Geochemistry data on Okanogan and Colville Forests (Boleneus and Chase, 1999a)

Procedure

The weights-of-evidence analysis conducted upon lithologic units can be generally described as occurring in three steps. First, a categorical (weights) analysis was used to select a specific group of lithologic units from the 169 lithologic units that are areally-associated with the training set. Second, a cumulative (weights) analysis was performed on the specific group of lithologic units to improve the predictor theme, and third, a cumulative analysis was performed on data consisting of acid-to-intermediate Eocene-age dikes. A cumulative analysis is appropriate in the latter two steps because buffer bands are constructed around the features.

Categorical analysis

The categorical analysis was used to select specific lithologic units nearest to the deposits. This procedure determines the areal association between the training set and the geologic units that occur at the surface. It is recognized that geologic units in the subsurface may often differ from surface units. Because of the areal association (i.e. proximity) with the training set, the following nine lithologic units were selected: Eck, Evkct, Evst, Evsf, Evkf, Eco, Eid, TRPMmsv, and Q (table 2a). A description of these units is given below. These lithologic units were selected based on calculated weights and contrast values. Where the contrast value is positive, the units are considered inside the pattern of prediction (“inside pattern”), if the contrast is 0.0 or less, the unit is outside the pattern (“outside pattern”). Where the contrast values range

from 0.01 to 1.0 the association is considered mild to moderate, where the contrast ranges from 1 to 2 the level of prediction is strong; where the contrast value is 2.0 or greater the association is considered extremely predictive. The areal extents of the lithologic units having positive contrast values clearly define the limits of the Republic and Toroda Creek grabens (fig. 4a). The remaining 160 units not selected are shown in Appendix I (table 2b). The Quaternary undivided (Q) unit is not shown in this figure.

The Eocene rock strata selected above consist of three successive units. The lowermost unit, the O'Brien Creek Formation (Eco), consists of varied pyroclastic and mudflow units formed by explosive quartz latite volcanism. The Sanpoil Volcanics consist of rhyodacitic lavas and breccias (Evsf, flows; Evst, tuffs and volcanic breccia). The uppermost unit, Klondike Mountain Formation consists of flows, breccias, and domes in addition to bedded tuffs and lake bed deposits (Eck, volcanoclastic rocks and sediments; Evkct, volcanic conglomerate and tuffs; Evkf, flows) according to Pearson and Obradovich (1977).

Six Eocene volcanic or sedimentary rock units are considered non-prospective for epithermal deposits and were not selected during the analysis. These include Evsv (volcanoclastic unit of the Sanpoil Volcanics in Nespelem and Chewelah quadrangles), Evf (unnamed flows), Evcg (unnamed volcanic conglomerate), Evcl (unnamed volcanoclastic rocks), Ev (unnamed volcanics undivided) and Ec (Tiger Formation near Newport, Washington).

Table 2(a). Proximity analysis for lithologic map units.

[Results are categorical and are ranked by contrast. The high inside-the-pattern areas (bold typeface) are separated from low inside-the-pattern area (italic typeface) at a contrast value of 2. Contrast values > 2 are considered extremely predictive. See Kemp and others, 1999 and App. II, Glossary. See Appendix I for Table 2b. Quaternary (Q) unit is shown because it contains 15 training sites; W⁺ and W⁻ are weights; Eck, Evkc, Evkf – Klondike Mountain; Evst, Evsf – Sanpoil Volcanics; Eco – O'Brien Creek Formation; Eid – Dikes (Eocene); TRPMmsv – metavolcanic rocks, undivided; Q – Quaternary]

Lithologic symbol	Class	Area, km ²	Number of training sites	W ⁺	W ⁻	Contrast	Rank
Eck	44	26.3	4	4.4686	-0.0825	4.5510	Inside pattern, high
Evkct	65	29.0	2	3.5844	-0.0397	3.6241	Inside pattern, high
Evst	75	17.2	1	3.4007	-0.0195	3.4202	Inside pattern, high
Evsf	21	656.2	20	2.7273	-0.4843	3.2116	Inside pattern, high
Evkf	61	212.9	5	2.4596	-0.0968	2.5563	Inside pattern, high
<i>Eco</i>	<i>20</i>	<i>70.6</i>	<i>1</i>	<i>1.9446</i>	<i>-0.0173</i>	<i>1.9619</i>	<i>Inside pattern, low</i>
<i>Eid</i>	<i>16</i>	<i>163.4</i>	<i>1</i>	<i>1.0968</i>	<i>-0.0135</i>	<i>1.1103</i>	<i>Inside pattern, low</i>
<i>TRPMmsv</i>	<i>9</i>	<i>176.5</i>	<i>1</i>	<i>1.0193</i>	<i>-0.0130</i>	<i>1.0323</i>	<i>Inside pattern, low</i>
Q	5	<u>9712.0</u>	<u>15</u>	-0.2844	0.1520	-0.4364	Outside pattern
		11061.0	50				

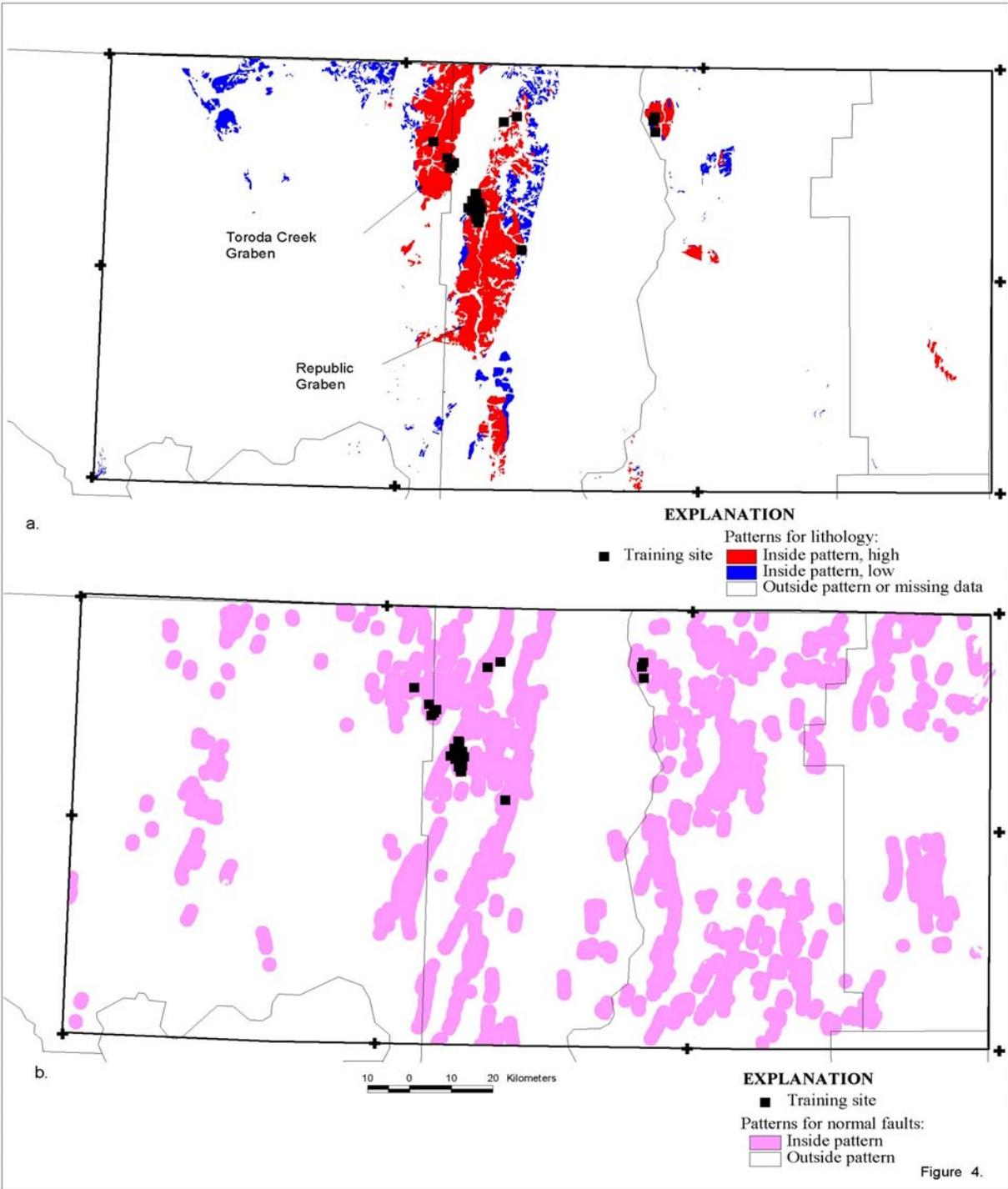


Figure 4.

Figure 4. Lithologic units pattern and normal faults pattern maps.

(a) Shows extent of pattern for lithology units selected by the categorical analysis. The “inside-the-pattern” areas clearly define the limits of the Toroda Creek and Republic grabens. (b) Predictor pattern of 1700 m radius surrounding all normal faults (buffer and faults are not shown). Training sites are shown.

Discussion of categorical data

The nine units consisting of Eck, Evkct, Evst, Evsf, Evkf, Eco, Eid, TRPMmsv were selected within the predictor theme because the contrast values > 0 . The group consisting of lithologic units of the Klondike Mountain Formation and Sanpoil Volcanics (bold typeface in table 2) received a contrast value > 2.0 and are assigned the inside the pattern, high ranking. This could be expected because gold deposits occur within the Sanpoil Volcanics. Three other units, consisting of O'Brien Creek Formation, Eocene dikes and meta-volcanics, undivided, received an inside the pattern, but lower rank (italic typeface) and their association with gold deposits is less clear. The Quaternary undivided (Q) unit lies outside the pattern and is considered an area of missing data because the composition of the underlying bedrock remains unknown. Since all lithologic units in the table (except Q) have high contrast values, they have strong to extremely predictive areal associations with the training set. From this association, it can be concluded that area covered by the eight lithologic units (selected units) have a greater likelihood for containing undiscovered epithermal gold deposits.

The numbers of training sites found inside the borders of each lithologic unit are shown in table 2a. The Evsf unit, consisting of flows of the Sanpoil Volcanics, contains the largest number of training sites. These results were anticipated since the best deposits in the Republic, Curlew and Orient areas occur near the top of the Sanpoil Volcanics just below the contact with the overlying Klondike Mountain Formation (Fifarek, Devlin, and Tschauder, 1996; Steven Box, written commun, 2001). The Evsf also occupies an area of flow units of the most common host (Muessig, 1967; Full and Grantham, 1968). As a group, the five units that rank *inside the pattern, high* are extremely predictive and contain 32 of the 50 training sites. Three other training sites occur *inside the pattern, low* in the O'Brien Creek Formation, Eocene dikes, and undivided metavolcanic rocks. These three units are all considered strongly predictive patterns. The 15 sites found with the Quaternary (Q) unit are troublesome and represent the principal reason to perform the cumulative analysis. The Q unit is problematic because it contains such a large number of the training sites with unknown bedrock composition below.

The categorical analysis of lithologic units provides a good predictor theme. The eight classes have been summarized to three classes, high-inside pattern and low-inside pattern, and outside pattern. Before carrying out the cumulative analysis for lithologic units, the results of the categorical analysis of lithologic units were run in the final model to observe that outcome. It is an advantage of weights-of-evidence that several scenarios can be tested in a short time. In the final model, one predictor theme of categorical data containing three classes was found to unnecessarily bias the results in favor of lithology⁴. The problem that the Quaternary unit contains 15 training sites was mentioned. Both of these problems were overcome by deriving a cumulative analysis of the selected lithological units for the final model. In the following discussion the categorical results (above) are re-analyzed using a cumulative analysis. The cumulative analysis results replaced the categorical results as an improved predictor theme of

⁴ An important assumption of weights-of-evidence analysis is that predictor themes must be independent (see conditional independence in Glossary) of one another. Results from both the categorical and cumulative analysis were separately integrated into final epithermal gold models and it was found that the cumulative analysis results provided the highest independence.

lithologic units. The cumulative analysis results are a better predictor theme because of higher contrast values and they account for 47 of the 50 training sites.

Cumulative analysis

The hypothesis, that one pattern containing the eight selected lithologic units is spatially correlated with the training set, was tested. In this step, the cumulative analysis was performed on the group of eight lithologic units (Eck, Evkct, Evst, Evsf, Evkf, Eco, Eid, TRPMmsv) from the categorical analysis. The eight units were grouped and analyzed as a single area. First, a series of buffer bands each of 50-m width were digitally constructed around the area before performing the analysis. The cumulative results (table 3) show the contrast values forms a peak at the 150 m buffer radius. This area encompasses 47 of the 50 training sites. The distance where the contrast is highest, forms a distinct peak, and accounts for the larger number of the training sites is the optimum distance. The optimum distance defines the limit of the predictor pattern (inside the pattern). All remaining areas, including the three missed training sites, are outside the pattern. Inside the pattern includes the area of all eight grouped units and extends outboard 150 m from their border. The peak contrast of 5.32 is an improvement over the best contrast of 4.55 in the categorical discussion. Its only class is inside the pattern (replaces three classes of categorical results). The conclusion is that the buffered pattern of lithologic units is an extremely good predictive theme based on the level of the contrast value. This theme will be used later in the data integration step as one of three themes to create the epithermal gold deposit model. A separate figure was not prepared of the results of the cumulative analysis since no differences can be observed at the scale of the map (fig. 4a).

Table 3. Proximity analysis for eight selected lithologic units.

[The optimum buffer distance at 150 m highlighted in bold typeface is the limit of “inside the pattern”(inside pattern). Buffer distance, area, and number of training sites are cumulative. At the limit of the 150-m radius pattern, it is noted that 47 of the 50 training sites occur within the optimal pattern. That area beyond 150 m is considered *outside the pattern* and contains three training sites.]

Buffer distance, m	Area, sq. km.	Number of training sites	W +	W -	Contrast
50	1493.8	40	2.59	-1.55	4.14
100	1630.8	45	2.62	-2.24	4.86
150	1776.8	47	2.58	-2.74	5.32
200	1890.6	50	--	--	--

Dikes theme

Small-scale, shallow intrusive features of intermediate to acid composition, or dikes, were expected to show the area of extensive dacitic volcanic activity of the epithermal event. Dikes having a basic or alkalic composition or those of other than Eocene age were excluded. Accordingly, such a group of dikes of Eocene age was assembled from the dikes coverage (Schuster and Harris, unpublished data) and tested as a possible predictor theme. Results from analysis showed the optimum contrast occurred at 9000 m. The conclusion was drawn that such a pattern had a weak yet positive spatial relationship to the training set. The pattern outlined the huge epithermal volcanic field centered in northern Ferry County. However, the results (not

shown) were not used as a predictive theme because they lacked geological significance, covered an extensive area, and some of the dikes were already included in the Eid lithologic unit of the previous categorical and cumulative analysis. Such a buffer seemed too wide, not a particularly useful unbiased predictor layer, and unlikely to provide a significant improvement of the final epithermal model. Another problem with dikes theme is that the small dikes are not universally included on 1:100,000-scale geologic maps and so all dikes may not be available for the analysis.

2. Proximity to normal faults

Epithermal deposits develop along faults and fractures that formed in response to the region's major structural adjustments during volcanic events (Full and Grantham, 1968). Increased permeability along faults probably controlled the pathways followed by fluids that deposited metals and gangue minerals. Therefore faults are considered as potential localizers for ore deposition. A hypothesis that all normal faults controlled mineralization was tested. The faults sub-coverage contained 5148 fault segments of all types and 4480 normal fault segments were selected for the analysis. These results (not shown) were rejected because the contrast value was not definitive. The conclusion was drawn that such a large group of normal faults was not spatially related to the training set.

A second hypothesis addressing control on mineralization by normal faults of a particular orientation was formulated and tested using a modified version of the normal fault coverage. In this version, faults were classified by compass orientation and type. The northerly-trending structural trends of normal faults are significant sites for deposition of gold. It is clear from the geologic map of northeast Washington (Stoffel and others, 1991) that major structural trends in Eocene rocks occur in or near the Republic graben along azimuths ranging from 10° to 20°. Vein azimuths in the Eureka fault trend near Republic range from 330° to 10° (Muessig, 1967). In a comparison of all faults occurring in all compass directions in the study area, the larger proportion of normal fault segments occur in the range from 345° to 30°. The trend of this test group is consistent with the dominant fault and vein direction of northeast to northwest (Full and Grantham, 1968). The selection excluded faults classified as thrust faults, low-angle normal faults, or those of unknown type. Of a total of 4480 normal fault segments, a group of 1807 normal faults occur within the 345° to 30° interval (table 4) and subjected to the hypothesis testing.

Table 4. Number of normal fault segments whose orientations range from 345° to 30°.

Fault azimuth group	Number of fault segments
345-350°	153
351-360°	358
1-10°	426
11-20°	459
21-30°	<u>411</u>
Total	1807

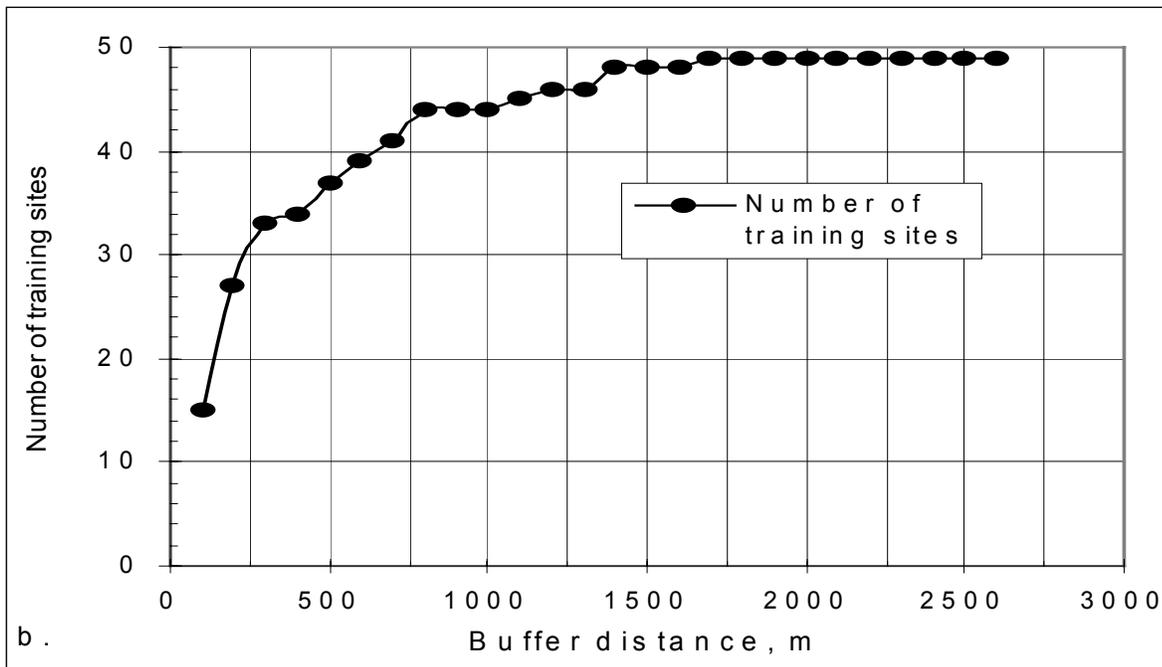
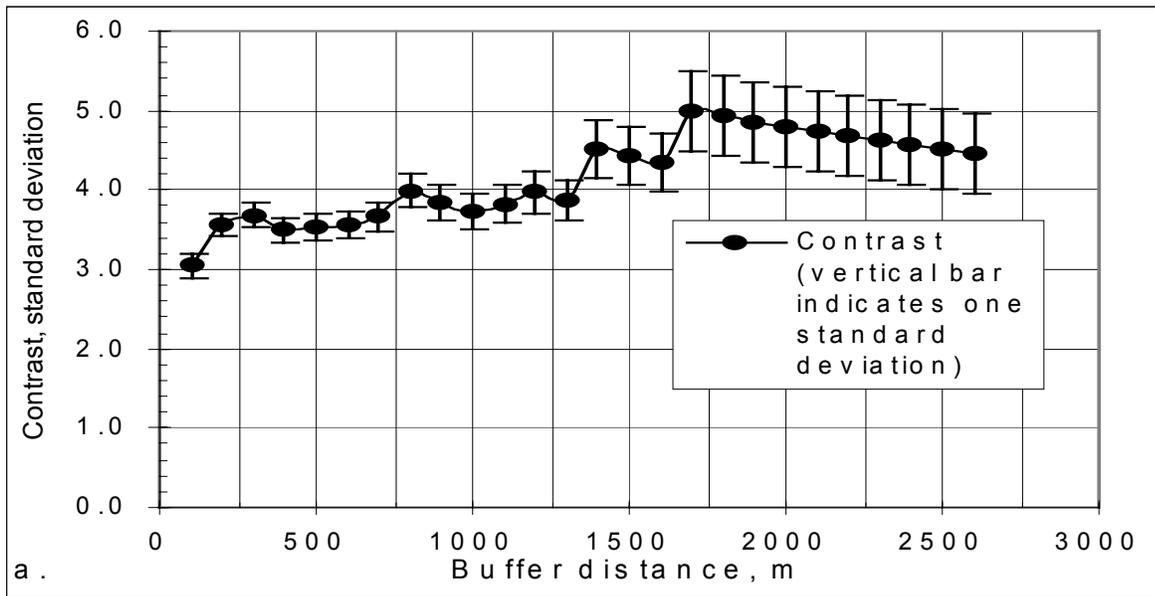


Figure 5. Normal faults pattern analysis.

(a) Contrast versus buffer distance for normal faults theme. Shows peak contrast occurring at 1700 m distance from each fault. Error bar represents one standard deviation. (b) Number of training sites versus buffer distance at successive distance from faults. Shows that 49 training sites are captured within the 1700-m buffer distance.

The 345°-to-30°-oriented normal fault group was buffered with 100 m wide bands extending in all directions from each fault to 2700 m. The cumulative proximity analysis results indicate that the contrast peaks at 1700 m (fig. 5a) at a contrast value of 4.99 (table 5). The faults pattern (inside pattern) area includes this particular group of normal faults and the area extending outboard to the 1700-m buffer. The pattern includes 49 of the 50 training sites (fig. 5b). The pattern is presented in generalized form (fig. 4b), as the normal fault segments are too numerous to show at the scale of the map. We conclude that the normal faults pattern is an extremely good predictive theme based on the contrast value. The pattern is used later as a predictor theme in the epithermal gold model.

Table 5. Proximity analysis for normal faults.

[The optimal buffer distance at 1700 m (bold typeface) represents the limit of “inside the pattern” for normal faults. It encloses 49 of 50 training sites. Buffer distance, area, and number of training sites are cumulative. The area outboard of 1700 m is “outside the pattern”.]

Buffer distance, m	Area, km ²	Number of training sites	W ⁺	W ⁻	Contrast
100	498.0	15	2.72	-0.34	3.05
200	813.9	27	2.81	-0.74	3.56
300	1161.9	33	2.65	-1.03	3.69
400	1500.5	34	2.42	-1.08	3.50
500	1897.5	37	2.27	-1.27	3.54
600	2243.5	39	2.15	-1.42	3.57
700	2582.6	41	2.06	-1.60	3.66
800	2949.6	44	2.00	-1.99	3.99
900	3337.2	44	1.87	-1.97	3.85
1000	3709.0	44	1.76	-1.96	3.72
1100	4043.8	45	1.70	-2.12	3.82
1200	4372.9	46	1.64	-2.33	3.97
1300	4771.0	46	1.56	-2.31	3.86
1400	5114.4	48	1.53	-2.99	4.51
1500	5468.0	48	1.46	-2.97	4.43
1600	5781.2	48	1.40	-2.95	4.35
1700	6121.9	49	1.37	-3.62	4.99
1800	6440.3	49	1.32	-3.61	4.92
1900	6772.2	49	1.27	-3.59	4.85
2000	7092.3	49	1.22	-3.57	4.79
2100	7375.1	49	1.18	-3.55	4.73
2200	7694.6	49	1.14	-3.53	4.67
2300	7977.1	49	1.10	-3.52	4.62
2400	8253.0	49	1.07	-3.50	4.57
2500	8560.4	49	1.03	-3.48	4.51
2600	8840.1	49	1.00	-3.46	4.46
2700	9111.9	50			

3. Proximity to placer gold sites

Several geochemical themes were evaluated as predictors of epithermal deposits. Among these, gold and silver analyses from stream sediment samples (Boleneus and Chase, 1999) were tested but were rejected as predictor themes. These were rejected because no optimum contrast

level could be determined and because very large buffer distances failed to capture a substantial number of training sites. We speculated that the difficulty in using the gold-silver geochemistry theme is related to incomplete coverage of the silt geochemistry sampling sites. The gold and silver silt sample sites were restricted to public lands in national forests while the majority of the training sites are found on private lands. These results are not shown. Gold and silver analytical values from NURE spring and stream water samples may be a reasonable geochemistry theme but it was not tested. The disadvantage in having two geochemistry themes is similar to the problem with two lithologic themes discussed above.

We anticipated that placer gold mines would be a positive indicator of proximity to gold mineralization because they could indicate eroding deposits somewhere upstream. Gold placers are a likely indicator because winter precipitation is heavy and spring stream flow is strong. A total of 67 placer sites were obtained from the U.S. Geological Survey MRDS and US Bureau of Mines MILS databases (Boleneus, 1999a). These include all placer sites available from within the study area. The only criteria applied were that either gold was reported in placer form or was mined from placer deposits (Appendix I).

Using the proximity analysis, the hypothesis was tested that placer gold sites spatially correlate with the training sites. The 1000-m-wide buffer bands (colored) were first drawn around each placer gold site (fig. 6a) and extend outward to a radius of 10,000 m. The bands were made in the same way for the lithologic units (not shown) and normal faults themes (not shown). Proximity analysis results indicate that the contrast peaks at 4000 m (fig. 7a) at a contrast value of 3.70 (table 6). The area inboard of the peak contrast at 4000 m encloses 43 of the 50 training sites (fig. 7b). Based on this analysis, the 4000 m buffer is chosen as the optimum distance of the predictor pattern. We conclude that the pattern (fig. 6b) for placer sites is an extremely good predictor pattern based on this contrast value. The generalized placer predictor theme is the third of three predictor themes in the epithermal gold model.

Discussion

There was some concern that the circular buffer pattern for placers may be misleading because they are not pointed upstream toward the source of the gold. However, the density of points was insufficient to distinguish any direction. A possible disadvantage is that an age or geologic unit association cannot be established between the placer gold and training sites. We found these concerns were negated in the final model by the interaction of the placer sites with the other predictor layers.

The seven of the 50 training sites occurring outside the placer sites pattern are a concern. The explanation may be related to discovery of hidden deposits, K-2 and Kettle. K-2 and Kettle mines are two examples of the seven excluded training sites. Echo Bay Inc. (M. Rasmussen and C. Gelber, verbal communication, 2000), operator of both mines, indicated that both deposits were eventually discovered after completing follow-up sampling of anomalous gold values found by extensive proprietary stream silt sampling in the nearby, east-west drainage. Deposits are hosted in Eocene Sanpoil Volcanics and found drilling beneath thick Quaternary overburden. The lack of placer gold near these hidden deposits may also be due to the concealing Quaternary overburden. This explanation eases the concern that 7 of the 50 occurred outside the pattern.

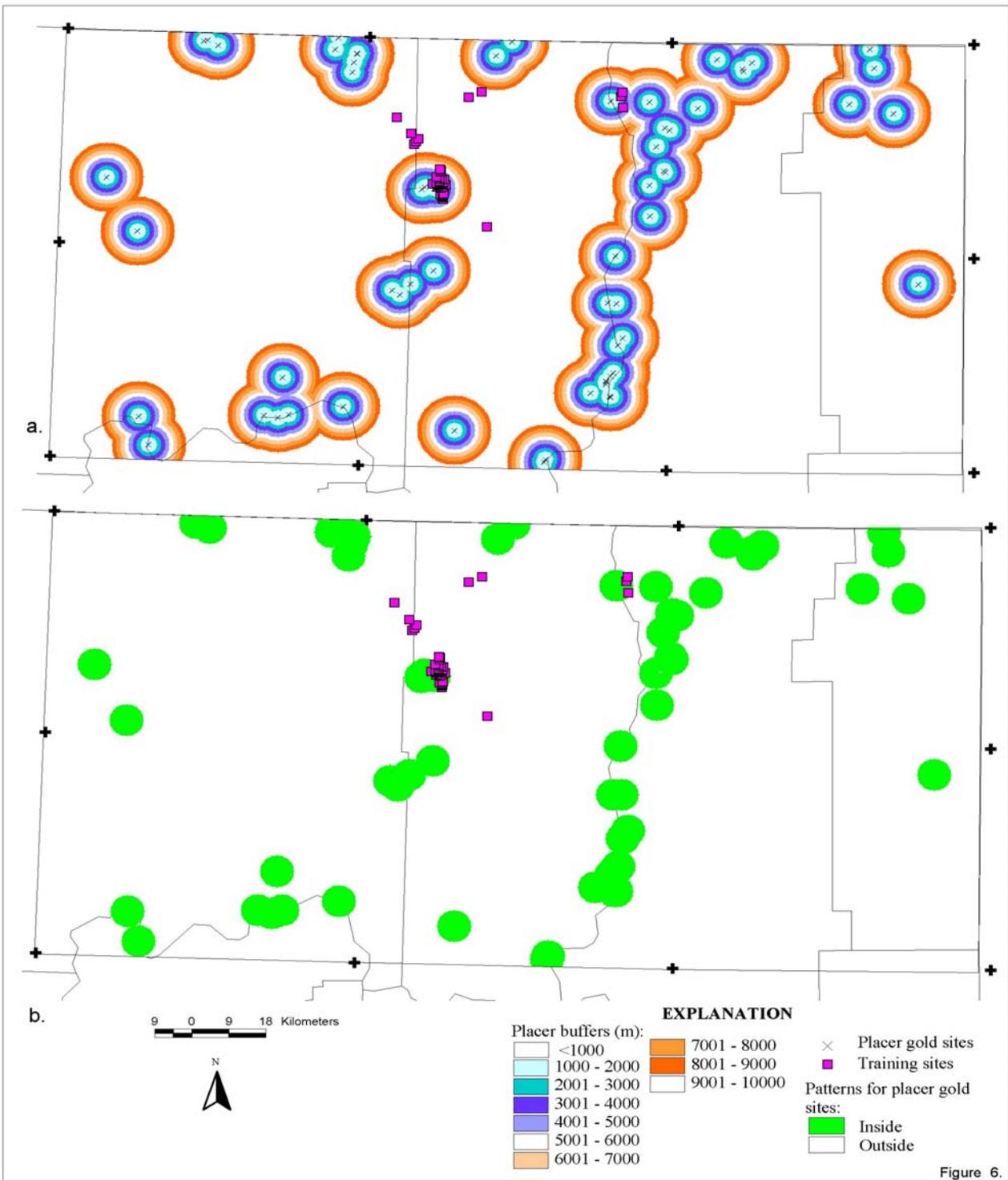


Figure 6.

Figure 6. Placer gold pattern map

(a) Ten, 1000-m wide buffer bands (multi-colored) created around placer sites before analysis. (b) Shows optimum 4000 m radius pattern as a result of the analysis. Placer sites occur at center of each circular pattern (not shown). Training sites are also shown.

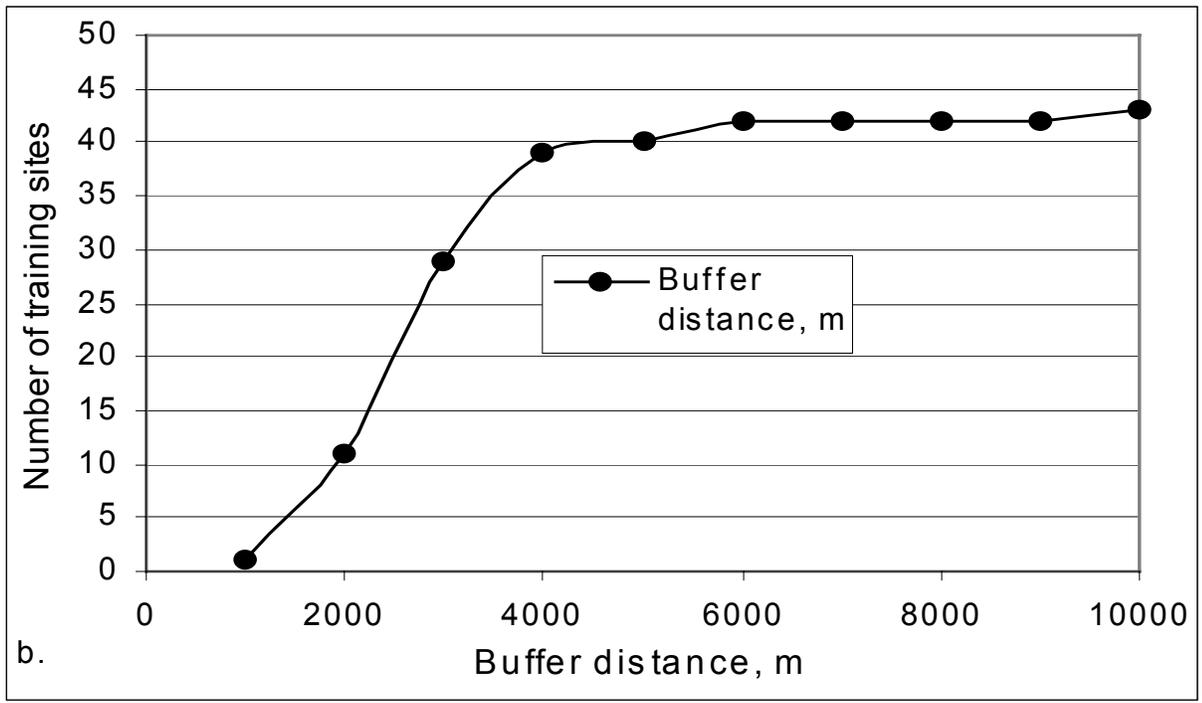
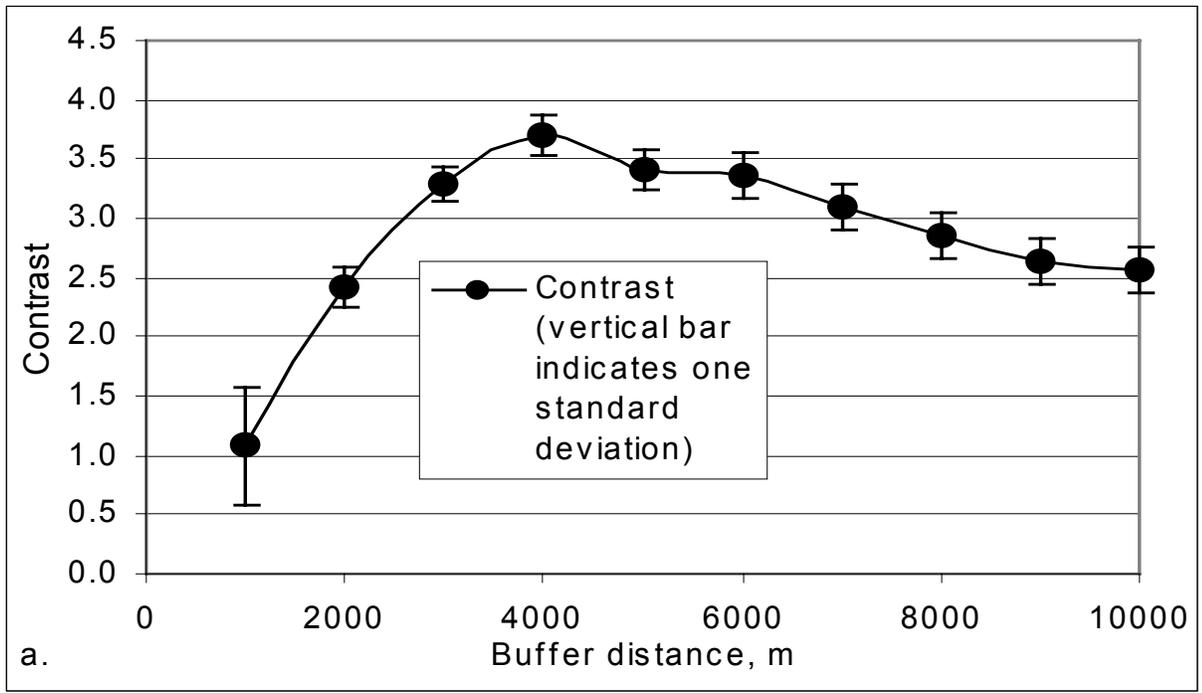


Figure 7. Placer gold pattern analysis.
 (a) Shows contrast (and standard deviation as a measure of error) versus buffer distance. Buffers are 1000m-wide colored bands constructed around each placer gold site shown in previous figure. (b) Shows cumulative number of training sites captured within successive buffer bands versus distance. The 4000-m distance corresponds to the optimum distance established above.

Table 6. Proximity analysis for placer gold sites.

[The optimal buffer distance at 4000 m (bold typeface) establishes the limit of “inside the pattern”. Buffer distance, area, and number of training sites are cumulative.]

Buffer distance, m	Area, km ²	Number of training sites	W ⁺	W ⁻	Contrast
1000	168.8	1	1.06	-0.01	1.08
2000	606.6	11	2.20	-0.22	2.42
3000	1223.0	29	2.47	-0.82	3.29
4000	2003.5	39	2.27	-1.43	3.70
5000	2880.7	40	1.92	-1.49	3.41
6000	3770.3	42	1.70	-1.67	3.37
7000	4739.9	42	1.47	-1.62	3.09
8000	5725.5	42	1.28	-1.57	2.85
9000	6729.0	42	1.12	-1.51	2.63
10000	7800.7	43	0.99	-1.58	2.57
>10000	24372.9	50			

4. Integrating patterns for epithermal gold model

The patterns for three predictor themes are combined to form the epithermal gold deposit model. Table 7 summarizes the weights-of-evidence data for this model following integration of the predictor theme patterns. The contrast values indicate the strength of association of themes with the training set. Theme 1 was obtained from the categorical analysis for lithology and is included here only for comparative purposes. Themes 2, 3, and 4 are those used in the model. The *lithology theme* (theme 2) is the strongest predictor theme as indicated by its extremely high contrast value of 11.0. Its very strong W⁻ of -8.4 (outside the pattern) indicates that unfavorable lithologic units are not associated with the training set. The high W⁺ of 2.5 (inside the pattern) indicates that lithology is an extremely strong targeting pattern. The *normal faults theme* and *placer gold theme* (themes 3 and 4) are similarly characterized, however, their contrast values differ from the lithology pattern.

The integration of all patterns produces the weights-of-evidence model for the occurrence of undiscovered epithermal gold deposits. In this process, digital patterns are laid one upon another because they can be registered by cell addresses. A completed model expresses the additive results of predictor themes as a posterior probability map presented in the next section. This is the probability that a unit cell in the grid contains a training point after consideration of all predictor themes. This measurement changes from cell to cell depending on the values in cells of each predictor theme, being larger than the prior probability⁵ where the sum of weights is positive. The posterior probability is the sum of the prior probability and the cell-by-cell weights of the predictor themes (Raines, Bonham-Carter, and Kemp, 2000).

⁵ Prior probability is the ratio of the number of training points to the area of the study area (50/23600 = .0021)

Table 7. Weights-of-evidence predictor theme data of epithermal gold model.

std. dev. – standard deviation							
Predictor theme	Criteria	Source	W ⁺	Std. dev. (W ⁺)	W ⁻	Std. dev. (W ⁻)	Contrast
1 Lithology (categorical)	Lithologic units selected: Evkct, Eck, Evkf, Evst, Evsf, Eco, Eid, TRPMmsv	Wash. DNR geology	ranges 1.1 to 2.7	ranges 0.58 to 0.18	-8.1	10	10.8
2 Lithology (buffered, cumulative)	0 to 150 m of lithologic units selected in theme 1 (lithology)	Wash. DNR geology	2.5	0.14	-8.4	10	11
3 Normal faults	0 to 1700 m of northwest-to-northeast-trending normal faults	Wash. DNR geology	1.4	0.14	-3.6	1	5
4 Placer gold	0 to 4000 m of placer sites	USGS MRDS and MAS-MILS	2.3	0.16	-1.4	0.3	3.7

Tract definition

Tracts must be defined before attempting to interpret the posterior probability map. *This paper adopts the approach that tracts may be defined as favorable, permissive, or non-permissive to contain undiscovered epithermal gold deposits based on the probabilistic (posterior probability and other measures) data that extend over the study area.* Defining the tracts involves an analysis of the geologic attributes that are characteristic for a given mineral deposit type. The foregoing validates the view that three predictor themes credibly express the characteristics of this deposit. The absence of a single critical characterizing attribute is sufficient to classify a tract as a “non-permissive” tract. In the past workers (Singer, 1993; Box and others 1996; Spanski, written communication) have defined tracts as permissive or favorable for the occurrence of a mineral deposit in question. The terms relate closely to the descriptive model for these deposits as we have already discussed. The *non-permissive* tracts are those that are believed to have virtually no potential or only negligible potential for the occurrence of a deposit (Singer, 1993). Singer suggests this condition might equate to a deposit occurrence probability of less than one chance in 100,000 or 1,000,000. By defining non-permissive tracts in this way, the remaining unexcluded tracts may be defined as being *permissive* for the deposit type sought. Spanski (written commun.) refers to exclusionary criteria as passive criteria for delineation of deposits. Permissive areas are those where the geology permits the existence of deposits based on geologic criteria derived from descriptive models about deposits. It is becoming commonplace in conducting assessments to designate a subset of the permissive tracts as *favorable* tracts based on active criteria (Spanski, written commun.) where tracts possess a significantly higher potential for deposit occurrence. The approach followed here resembles this method, as follows: Low to mid-range probabilistic data define the non-permissive area (exclusionary criteria), high probabilities define the permissive terrane, and highest probabilities (active criteria) define the favorable tracts.

The posterior probability values, P (exclusionary and active criteria) essentially define the limits of the favorable, permissive, and non-permissive tracts for the epithermal gold model. The

remaining probabilistic data from the weights-of-evidence analysis substantiate this conclusion. Table 8 shows the relationships of P to tracts following the integration step. The table contains eight classes corresponding to the number of possible combinations (inside, outside) for three predictor themes. P increases with number of training sites and sum of weights through the table. As viewed on curve on Figure 8, the P versus cumulative percent of the study area (fig. 8) provides a simple evaluation tool as the basis for separating the tracts on the posterior probability map (fig. 9). Table 8 provides the data to support this curve.

Criteria used in defining the favorable tracts (active criteria) are based on:

- Regions separated by natural breaks (inflection point) on the curve,
- Maintaining a consistent relationship between prior probability, p , (the known probability of the training sites) and the posterior probability, P ,
- Classes containing higher posterior probability values,
- Classes containing a larger number of training sites, and
- Classes containing predictor themes with highest contrast values,

The general method followed in applying these criteria is to limit the favorable area to a very small percentage (4.6%) and the non-permissive area to a very large percentage of the study area. For permissive tracts, the same measures are applied, although at lowered limits. The non-permissive tracts are those remaining after selection of the favorable and permissive tracts. The *prior probability* (0.0021) assists in defining tracts. Favorable tracts have posterior probability values greater than the prior probability, p , the non-permissive tracts have posterior probability values less than p , and the permissive area lies somewhere in the middle ground.

Notable features on figure 8 are positions of curve inflection points that lie just below posterior probability values of 0.024 and 0.00016. These prominent points tentatively establish the lower limits of the favorable and permissive areas, respectively (table 8) pending evaluation of other criteria. The non-permissive region falls below the lower inflection in the curve and represents 92 percent of the study area. The favorable region for this model, or 4.6 percent of the study area, lies above the upper inflection in the curve. The permissive region lies in the mid-range area and represents 3.4 percent of the study area.

From the foregoing, classes 1 through 4 (table 8) define the non-permissive area. Classes 5 and 6 define the permissive area and contain the one remaining training site, Kettle. Classes 7 and 8 define the favorable area with the highest weights and highest posterior probabilities. The region of classes 7 and 8 includes the area with 49 or 50 of the training sites. At the upper end of the range, class 8 is most prospective and represents 0.6 percent (154 km²) of the map area. Class 8 corresponds to that area common to “inside the pattern” to all three predictor patterns; class 8 also has the highest posterior probability for containing undiscovered epithermal gold deposits.

The three-fold division can be further characterized by the coinciding relationships of predictor themes. The non-permissive tracts lack the lithology predictor pattern, a fatal flaw. The permissive tracts contain the lithology pattern, at a minimum; class 6 contains placer sites as an additional theme. The favorable tracts contain both the lithology and normal faults patterns, at minimum; class 8 contains all three patterns.

Due to the uncertainty in this model, the application of posterior probability values should not be strictly applied but should be limited to ranking tracts as non-permissive, permissive, or favorable. That is, it is unwise to apply the posterior probability values as actual probability values to express the potential for undiscovered deposits.

Posterior probability map

The posterior probability map resulting from the integration process is shown in Figure 9. The colored areas represent the non-permissive (uncolored), permissive (blue), and favorable (red) tracts of the epithermal gold deposit model. The favorable and permissive areas largely fall within the area of the Republic and Toroda Creek Grabens corresponding to outcrops of Eocene volcanogenic rocks (compare fig. 4a and fig. 9). The uncolored areas are non-permissive tracts.

Table 8. Unique conditions data for epithermal gold deposit model.

⁽¹⁾ Buffered or cumulative weights theme for lithology ⁽²⁾ Sum of weights is defined as the sum of weights of contributing themes from Table 7 (for example, class 7 equals sum of lithology [W+], normal faults [W+], and placer sites [W-], or 2.5 + 1.4 +[-1.4] = 2.5)

Class	Tract definition	Patterns			Training sites	Area, km ²	Cumulative percent area	Posterior Probability	Sum of weights ⁽²⁾	Uncertainty
		Lithology ⁽¹⁾	Normal faults	Placer gold sites						
1		Outside	Outside	Outside	0	16,297	66.9%	0.00000001	-13.49	0.00000003
2	Non-permissive	Outside	Outside	Inside	0	1103	71.4%	0.00000011	-9.79	0.00000115
3		Outside	Inside	Outside	0	4393	89.5%	0.00000042	-8.49	0.00000419
4		Outside	Inside	Inside	0	609	92.0%	0.00001691	-4.80	0.00016917
5	Permissive	Inside	Outside	Outside	1	695	94.8%	0.00016703	-2.51	0.00017765
6		Inside	Outside	Inside	0	139	95.4%	0.00669556	1.18	0.00686940
7	Favorable	Inside	Inside	Outside	10	970	99.4%	0.02400974	2.48	0.00914021
8		Inside	Inside	Inside	39	154	100.0%	0.49815175	6.18	0.07388282
					50	24,359				

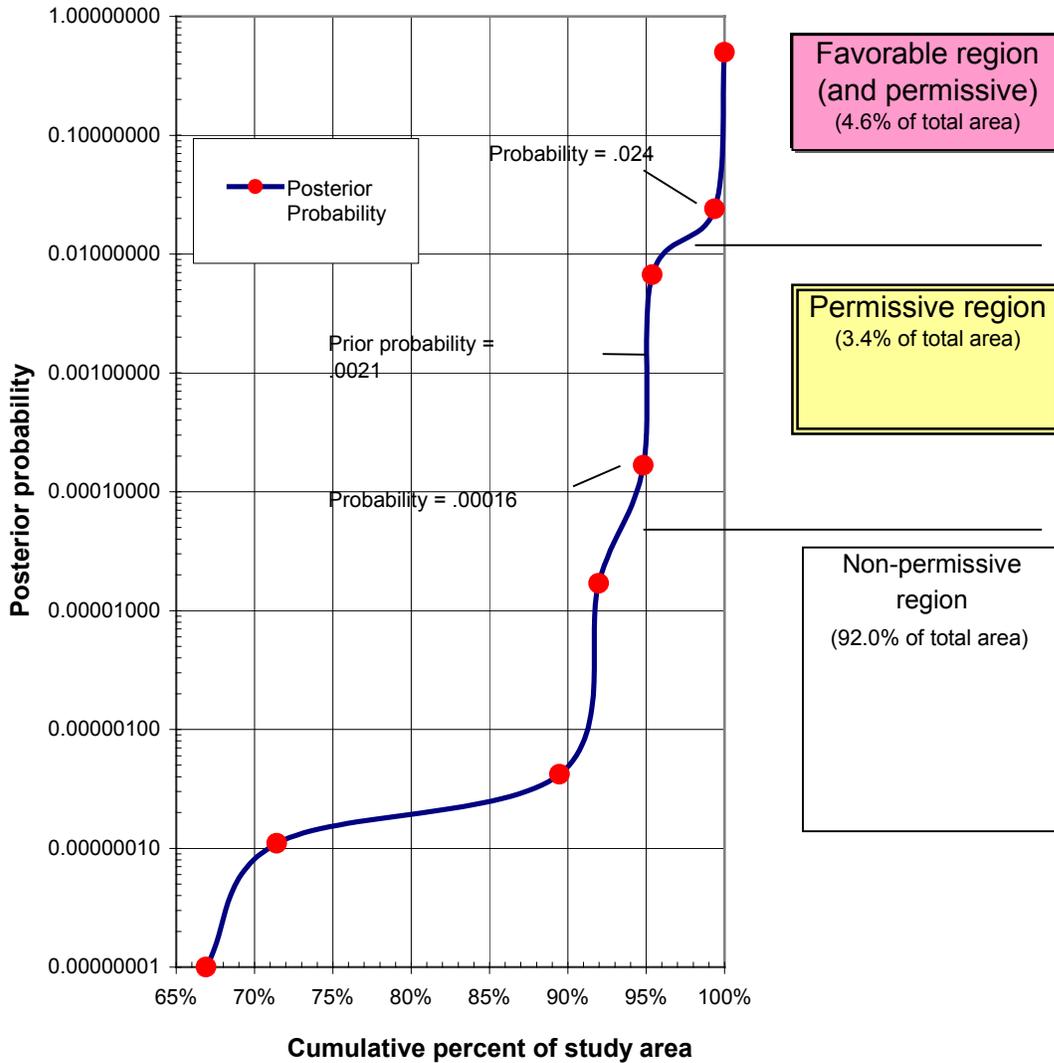


Figure 8. Probability results for epithermal gold deposit model.

Chart of posterior probability versus cumulative percent of study area for all unique conditions. The lower threshold of favorable region is assigned at 0.024 and includes 49 of 50 training sites. The lower threshold for the permissive region is arbitrarily assigned to 0.00016 and defines the class containing the prior probability. The prior probability is the known probability that the training set occurs in the study area. The permissive region also contains one training site, Kettle.

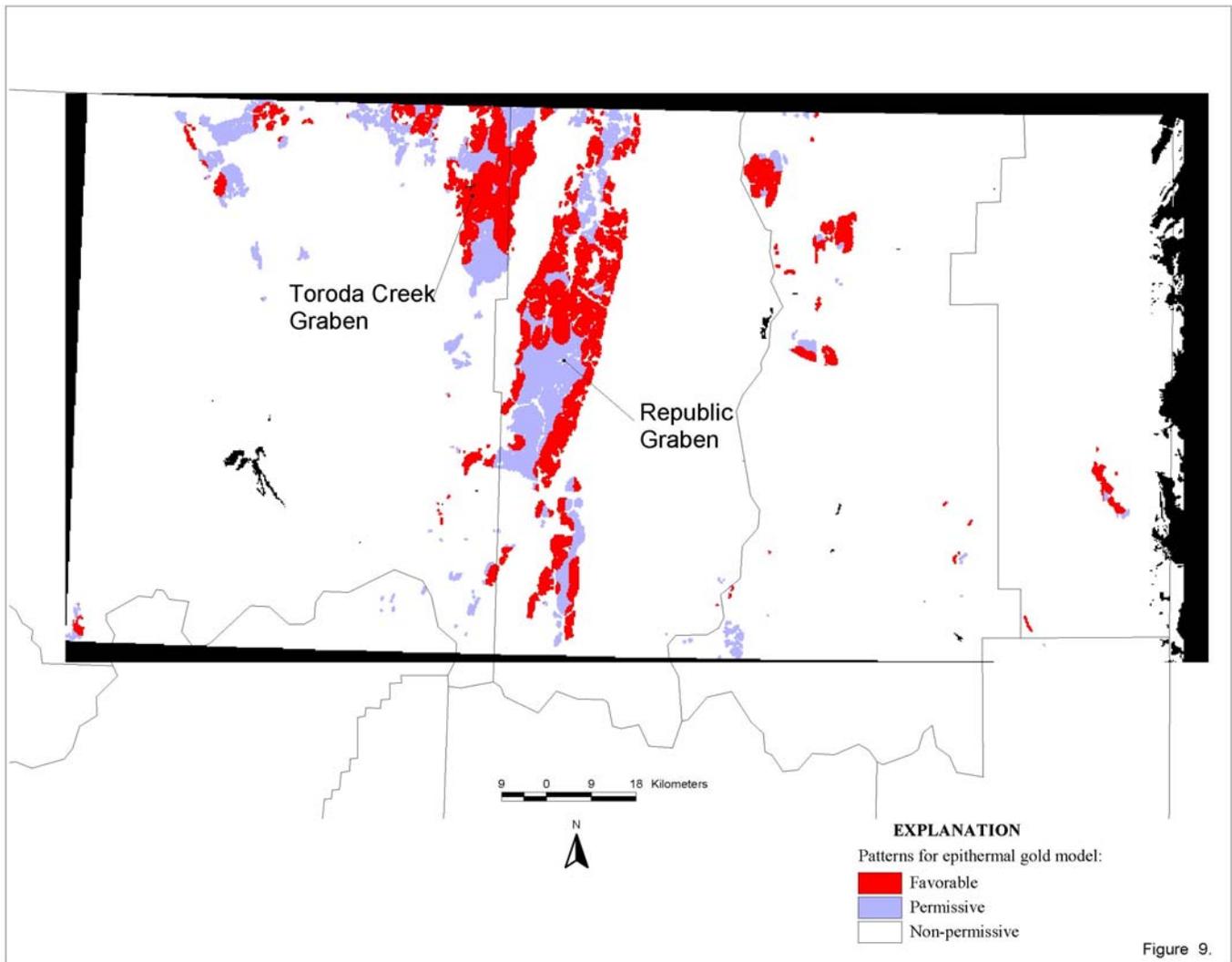


Figure 9.

Figure 9. Map of epithermal gold deposit model showing favorable, permissive, and non-permissive areas.

Red areas indicate favorable, blue areas indicate permissive, and uncolored area is non-permissive for discovery of undiscovered epithermal gold deposits. Black areas are artifacts of selecting a projected grid using a rectangular selection area.

MEASURES OF MINERAL ACTIVITY

Mining claims are used to measure the level of industry exploration activity for epithermal gold deposits. Other activity data such as mineral permits on national forests are also appropriate to measure activity but are not used here. The digital mine claim records supplied by the Bureau of Land Management are utilized for the period of 1980 to 1996. Activity is assessed based on “snapshots” of mining claim records at the end of years 1980, 1985, 1990, and 1996 (Hyndman and Campbell, 1999). The choice of years was arbitrary. Digital data about mining claims became available after 1978 when BLM started to maintain digital records. The terms historic, current and consistent are introduced here to rank claimed areas for their relative importance to predict future activity.

Historic activity is defined as a section (one square mile) containing either (a) one or more mining claims in 1980 or (b) one or more valid mining claims in 1985 (fig. 10). Current activity is defined as having one or more valid lode mining claims in the section during either 1990 or 1996. Consistent activity is section qualifying for both historic and current activity (that is, at least one section for 1980 or 1985 and at least one section for 1990 or 1996). “None” indicates no claims were present or that the land may not be open to locating claims.

Figure 10 indicates the outline of the Republic and Toroda Creek grabens containing a checkerboard pattern of claimed sections. It also shows that the area of consistent claims activity (red) is widespread, although less extensive than the extent of historic activity shown in green. There are pockets of consistent activity located within the Republic Graben and along the Washington-British Columbia border. Color shading of a section indicates that at least one claim is present in the section. One claim occupies an area of 20.7 acres (8.5 ha).

There are disadvantages in using mining claims as an activity measure. It is known that the area of interest for gold in the vicinity of historic mines is largely private land. So, on private lands, the mining claims approach to predicting activity is not valid. Understandably, mining claims are silent as to the type of locatable mineral and mineral deposit sought. Since the purpose of this study was to provide an evaluation of epithermal gold on the Colville and Okanogan National Forests, the method is a reasonable approach if confined within the region containing Eocene volcanoclastic rocks. We looked at the nature of exploration based on annual reporting of activity to the Washington Department of Natural Resources over the 13-year period from 1985 through 1997 within the Republic, Toroda Creek, Keller, and First Thought grabens. These areas contain over 98% of Eocene volcanoclastic rocks. Exploration was conducted on 71 projects during the period and all were conducted for purposes of gold or gold-silver exploration or development. However, on 23 (32%) of the 71 projects gold was sought in other than epithermal settings (Boleneus and Derkey, 2000).

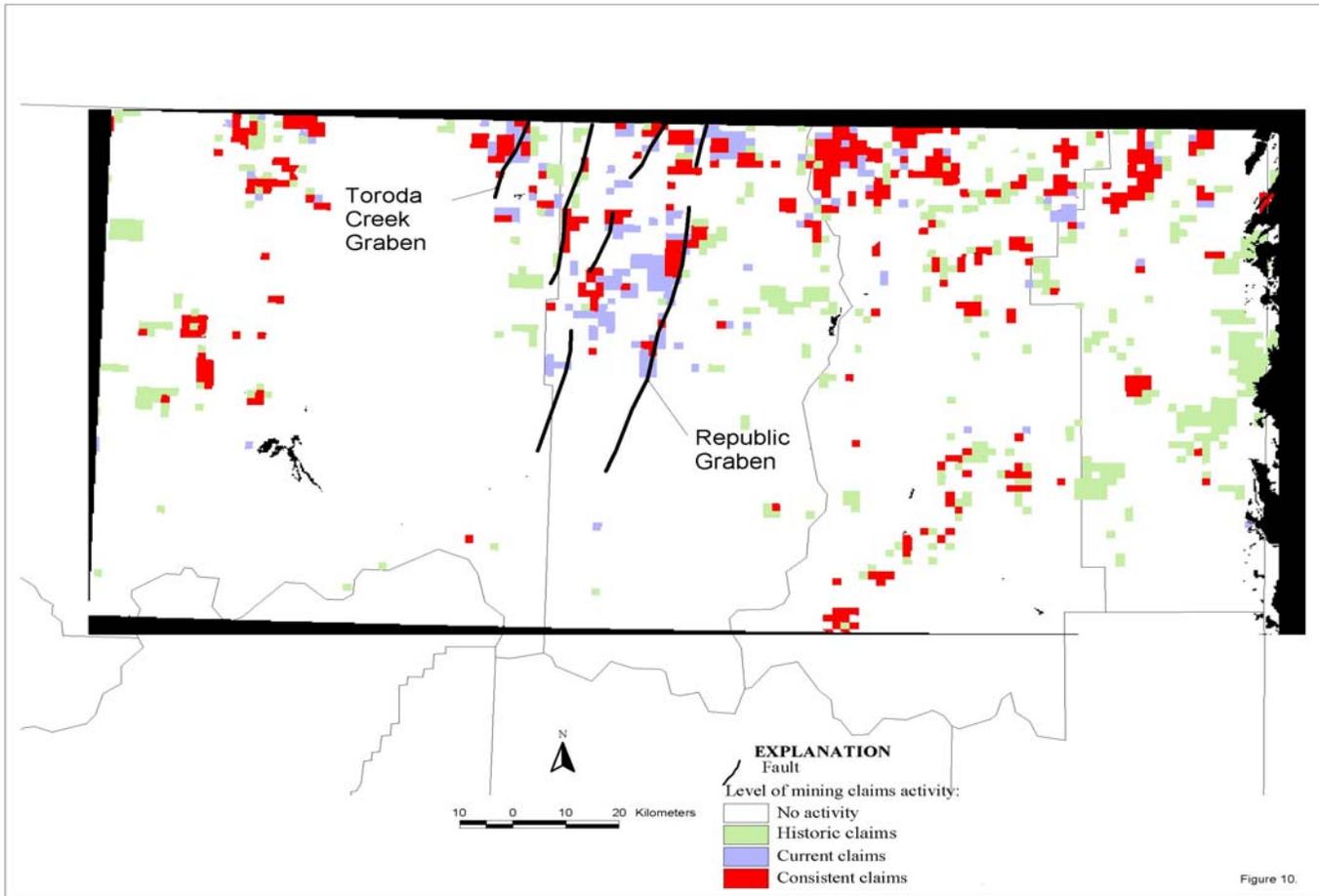


Figure 10.

Figure 10. Historic, current, and consistent mining claims activity.

Color shading indicates one or more valid mining claim exists on a section during the period. Green represents a section containing historic claims. Purple indicates current claims. Red indicates consistent (both historic and current) claims. See text for explanation of these terms.

ASSESSMENT OF FUTURE MINERAL INDUSTRY ACTIVITY

1. Mineral resource-activity matrix

A method is introduced here to relate the mining claims activity on federal lands to the deposit model for epithermal gold deposits. The purpose in investigating this relationship is to apply it as a land-management tool for federal land managers. The epithermal gold model is used as a measure of the undiscovered mineral resources and the mining claims are used as the measure of mineral industry exploration activity. The combination of the two data sets as called the *assessment model* for future exploration activity for gold deposits. The first step is to create a matrix to relate the two models, the mineral resource-activity matrix (fig. 11).

Taken together, the analysis of exploration activity and mining claim data are useful both as a predictor of future activity in seeking gold deposits and as a land management tool by federal land managers to assess the importance of minerals exploration activity on public lands. It is inconsequential to land managers that 32% of gold exploration is not for epithermal gold deposits. We assume that interest in exploration for epithermal gold deposits in the future will continue in areas known for past exploration activity.

The epithermal gold model has been divided into favorable, permissive, and non-permissive categories. The mining claim activity model has been grouped into four categories of activity defined as historic, current, consistent, and none. These classes are combined into the matrix with five categories that address the interrelated nature of both models. The matrix communicates the potential for future mineral exploration activity of epithermal gold deposits, ranked as follows: no activity predicted, low activity, low to moderate activity, moderate activity, and high activity. The ranks are relative terms that relate the likelihood of an undiscovered deposit to exist in an area in relation to the historic and current levels of claim activity. We chose to use these five categories because we believe that the potential nine categories in this matrix subdivided the information more than the quality of the information warranted. The application of the matrix is, of course, specific to the deposit type being sought. Use of this assessment method enables forest plans or other land-use planning on public lands to incorporate mineral-related information where necessary.

2. Assessment model

The resource model-activity model matrix is employed to create the assessment model (**fig. 12**). The assessment model represents the level of mining claims activity on public lands modified by the resource model (epithermal gold model). Supporting data about the assessment model are provided in the unique conditions table (table 9). The *high activity* category (red) includes favorable areas having consistent or current mining claims activity. The *moderate activity* category (yellow) is defined as favorable areas with historic or no mining claims activity. The *low-to-moderate* (blue) category consists of permissive areas with consistent, current, or historic claims activity. The little or no activity category (low, green) consists of permissive areas with no activity predicted.

EPITHERMAL GOLD MODEL	$p \geq .024$	FAVORABLE	POTENTIAL FOR HIGH ACTIVITY		HIGHEST ACTIVITY EXPECTED		WITH- DRAWN or NON- FEDERAL MINERAL ESTATE
	$p \geq 0.000167$	PERMISSIVE	LITTLE OR NO ACTIVITY	LOW TO MODERATE ACTIVITY			
		NON-PERMISSIVE	NO ACTIVITY PREDICTED				
			NONE	HISTORIC	CURRENT	CONSISTENT	
ACTIVITY MODEL							

Figure 11. Resource model-activity model matrix.

The matrix defines the interrelationship of the epithermal gold model and the mining claim activity model. Posterior probability (P) thresholds for the epithermal gold model are used to separate favorable, permissive, and non-permissive areas. Historic claims activity occurs on lands having valid claims in 1980 or 1985. Current claim activity occurs on lands having valid lode claims in 1990 or 1995. Consistent claims activity occurs on lands having land containing both current and historic claims.

Discussion

The locations of training sites are compared in the unique conditions of the assessment model (compare table 9, fig. 12). Seventeen of the training sites (classes 1, 5 in table 9) are located outside areas of current or historic activity and so these are likely located on private land holdings. Sixteen (class 5) of these sites are located within the favorable area based on the epithermal gold model. One site is located in class 1 that has little or no activity predicted. This site is the Kettle gold mine that is hosted by Eocene volcanic rocks. It occurs on private land but the deposit was buried beneath Quaternary deposits. About two-thirds, or 33 training sites (class 8) occur in the favorable epithermal gold model area associated with consistent mining claims activity. This indicates that about two-thirds of the favorable areas occur on lands that are both subject to location of mining claims and are of highest interest.

In comparing figures 9 (epithermal gold model) and 12 (assessment model) the favorable areas for the epithermal gold deposit model are subdivided into moderate activity and high activity areas on the map for the assessment model. That is, the yellow (class 5,6) and red (class 7,8) areas on the assessment map occupy the same as the red areas on the epithermal gold model map. Likewise the green (class -1,0) and blue (class 1) areas on the assessment model map are subdivisions of the purple areas on the epithermal gold model map.

Table 9. Unique conditions data for assessment model.

Mining claims activity		Epithermal gold model	Assessment model		Training sites	Area, km ²
Historic	Current		Class	Category		
Outside	Outside	--	-1	Unknown	0	12
Inside	Outside	--	-1	Unknown	0	1
Inside	Inside	--	-1	Unknown	0	<1
Outside	Inside	--	-1	Unknown	0	<1
--	--	--	-1	Unknown	0	<1
Outside	Outside	Non-permissive	0	No activity predicted	0	20,187
Inside	Outside	Non-permissive	0	No activity predicted	0	1,096
--	--	Non-permissive	0	No activity predicted	0	3
Inside	Inside	Non-permissive	0	No activity predicted	0	834
Outside	Inside	Non-permissive	0	No activity predicted	0	281
Outside	Outside	Permissive	1	Little or no activity	1	705
Inside	Outside	Permissive	2	Low to moderate activity	0	47
Outside	Inside	Permissive	3	Low to moderate activity	0	38
Inside	Inside	Permissive	4	Low to moderate activity	0	44
Outside	Outside	Favorable	5	Moderate activity	16	783
Inside	Outside	Favorable	6	Moderate activity	0	57
Outside	Inside	Favorable	7	High activity	0	143
Inside	Inside	Favorable	8	High activity	33	141
					50	24,373

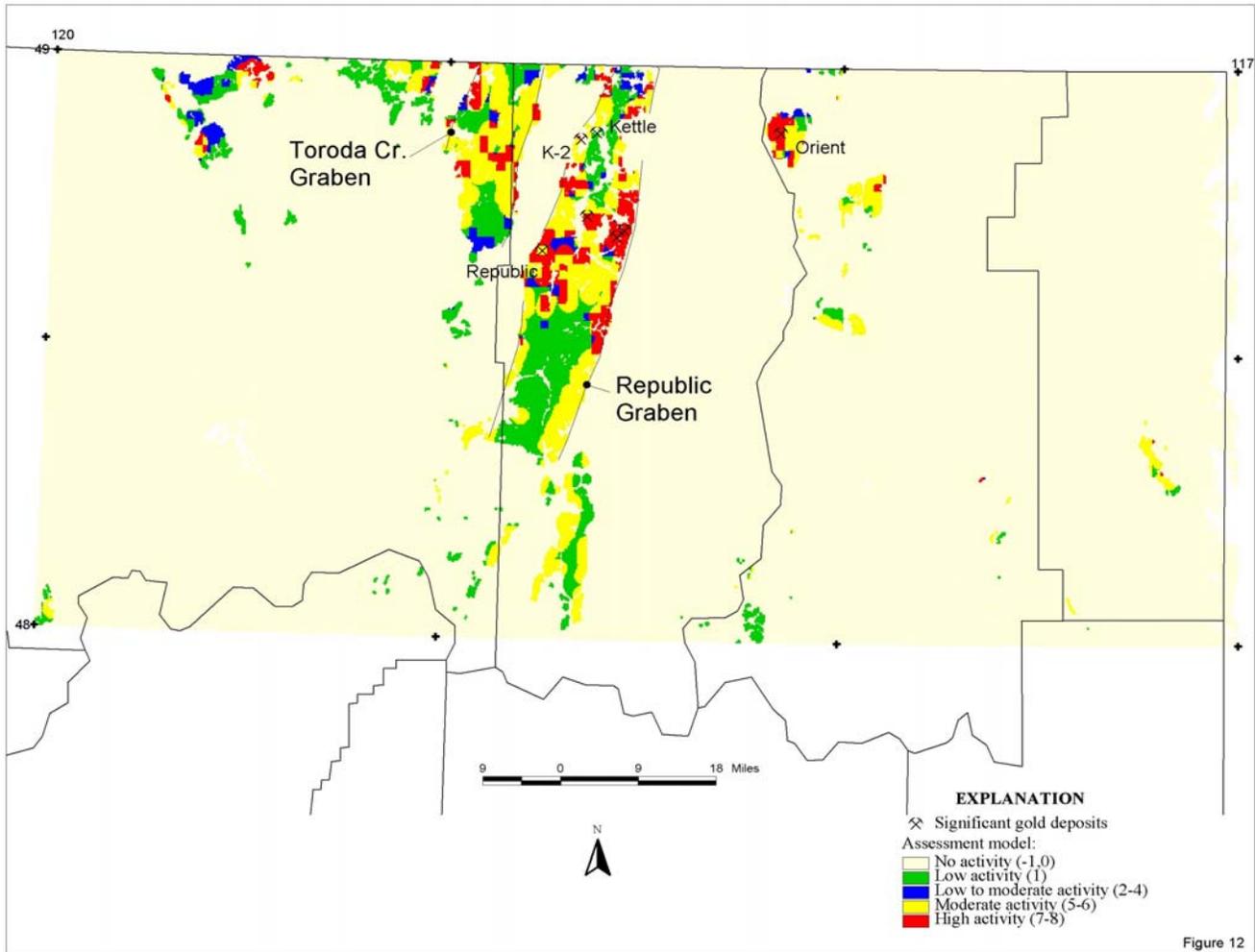


Figure 12

Figure 12. Map of assessment model.

This model consists of combined models for mining claim activity and epithermal gold. Numbers in parentheses refer classes defined in the table. Mine symbols that are named are significant epithermal gold deposits (Long, DeYoung, and Ludington, 1998). Unlabeled mine symbols are gold deposits of another type.

CONCLUSIONS

Hypothesis testing methods of the weights-of-evidence analysis were carried out on a number of specifically prepared digital geological themes for the evaluation of epithermal gold deposits in a 222 km x 277 km area of northeastern Washington State. A *training set* consists of 50 epithermal gold mines and prospects. A geologic map theme contained sub-sets consisting of lithologic units, faults, and folds. Geochemistry data included 67 placer gold sites.

The three themes describe the spatial correlation with the training set and are selected as the *predictor themes* during the analysis. They include the lithologic units theme, NW-to-NNE normal faults theme, and placer gold sites theme. The *lithologic units* pattern includes eight map units surrounded by a 150-m buffer. This theme areally describes the Klondike Mountain Formation and Sanpoil Volcanics, with known associations with epithermal gold deposits. The *normal faults theme* is a sub-set of normal faults having orientations that ranging from 345°-30°. This pattern includes the NW-NNE-oriented normal faults surrounded by a 1700-m buffer. The pattern for *placer gold sites* includes these sites surrounded by a 4000-m buffer. The epithermal gold model is formed by the integration (or overlayment) of these predictor themes. The probabilistic results of this model (1) describe the likelihood for occurrence of epithermal gold deposits and (2) form the basis for subdividing tracts into favorable, permissive, and non-permissive tracts according the grid cell-by grid cell presence of each predictor theme on the map. *Favorable* tracts were limited by posterior probability values (P) >0.024. This area also contained 49 of 50 training sites and covers 4.6 percent of the study area. The favorable area is restricted to the Republic, Toroda Creek, and Keller grabens, or areas of outcropping Eocene Klondike Mountain Formation and Sanpoil Volcanics. *Permissive* tracts represent another 3.4 percent of the study area where P>0.00016. *Non-permissive* tracts (92 percent) had P<0.00016. Other probabilistic data (patterns, *p*, sum of weights) assisted to define tracts.

Mining claim activity is defined by the locations of lode mining claims summarized for the years, 1980, 1985, 1990, and 1996. Terms defined here, historic, current, or consistent parcels, defined longevity and occurrence of mining claims during combinations of these years.

A matrix describing the interaction of the resource model-mining claim activity was defined to accommodate both the mining claim activity (historic, current, and consistent) and the occurrence for the epithermal gold deposits (favorable, permissive, and non-permissive). This combination describes the *assessment model* for gold deposits on public lands.

The assessment model represents the level of mining claims activity on public lands modified by the resource model (epithermal gold model) and assists federal land managers in land use decision-making by providing a prediction of mineral exploration activity in the next decade. The first category is *high activity* category which includes favorable tracts having consistent or current mining claims activity. The *moderate activity* category is defined as favorable tracts with historic or no mining claims activity. The *low-to-moderate* category consists of permissive tracts with consistent, current, or historic claims activity. The *little or no activity* category is permissive tracts with no activity predicted. The assessment model cannot be applied to lands other than public lands since mining claims are only staked on public lands. The model cannot be strictly considered for epithermal gold deposits because 32 percent (23 of 71 projects) of exploration and development projects occurring in the area of Republic, Toroda Creek, and Keller grabens during the 1985 through 1997 targeted gold in other than epithermal gold settings.

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APPENDIX

Appendix I. Databases

A. Training set for epithermal gold

Site name	MRDS Identifier number	MAS Identifier number.	Latitude	Longitude	USGS Model ¹	Production size	Posterior probability
1 Admiral	M056000		48.6600	-118.7369	25a	No production	0.0932
2 Advance	M056001		48.6267	-118.7453	25a	No production	0.0932
3 Alpine	M056023		48.6767	-118.7578	25a	No production	0.0932
4 Anecia	M056024		48.6619	-118.7803	25a	No production	0.0932
5 Ben Hur	SP00062		48.6686	-118.7581	25c	Small	0.0932
6 Blacktail (Hope)	SP00063		48.6633	-118.7475	25c	Small	0.0932
7 Bodie	M056027		48.6606	-118.7572	25a	No production	0.0932
8 Cook	M056029		48.6372	-118.7536	25a	No production	0.0932
9 East Sanpoil	M056030		48.6614	-118.7572	25a	No production	0.0932
10 El Caliph	M056031, SP00068		48.6592	-118.7639	25c	Small	0.0932
11 Flag Hill	M056032, SP00069		48.6522	-118.7536	25c	No production	0.0932
12 Golden Eagle	M056004		48.6600	-118.7369	25a	No production	0.0932
13 Ida May	SP00075		48.6547	-118.7647	25c	No production	0.0932
14 Insurgent	SP00076		48.6669	-118.7461	25c	Small	0.0932
15 Iron Mask	M056033		48.6561	-118.7544	25a	No production	0.0932
16 Iron Mountain	M056123		48.5639	-118.5994	25a	No production	0.0932
17 Jim Blaine Fraction	M056009		48.6328	-118.7453	25a	No production	0.0932
18 K2 Mine ²	na	0530190470	48.8660	-118.6680	25a	Large	0.0932
19 Kangaroo	M056034		48.6614	-118.7572	25a	No production	0.0932
20 Kettle	SP00080		48.8789	-118.6256	25a	Large	0.0007
21 Knob Hill mine and plant	na	0530190084	48.6734	-118.7578	25c	Large	0.0932
22 Last Chance	M056010, D001629, SP0008		48.6658	-118.7458	25a	Small	0.0932
23 Little Cove	M056036, SP00085		48.6669	-118.7533	25c	Small	0.0932
24 Lone Pine	M056037, SP00086		48.6653	-118.7506	25c	Moderate	0.0932
25 Mammoth	M056011		48.6725	-118.7428	25a	No production	0.0932
26 Mountain Lion	M056039 SP00095		48.6789	-118.7686	25c	Small	0.0932
27 North Sanpoil	SP00096, M056012		48.6667	-118.7586	25c	Small	0.0932
28 Old Hickory	SP00097, M056013		48.6392	-118.7428	25c	No production	0.0932
29 Pearl	M056040, SP00100		48.6669	-118.7533	25c	No production	0.0932
30 Princess Maude (Southern Republic)	SP00102		48.6347	-118.7475	25c	Small	0.0932

A. Training set for epithermal gold (cont.)

31 Quilp (Imperator, Eureka)	M056016	48.6561	-118.7475	25c	Small	0.0932
32 Rebate	M056041	48.6939	-118.7542	25a	No production	0.0932
33 Republic (Blaine Republic)	SP00104	48.6375	-118.7453	25c	Small	0.0932
34 Sanpoil Fraction	M056043, M056042, SP0010	48.6669	-118.7533	25a	Small	0.0932
35 Seattle	M056044, SP00107	48.6642	-118.7669	25c	No production	0.0932
36 Snowstorm	M056045	48.6614	-118.7572	25a	No production	0.0932
37 South Penn	SP00112, M056046	48.6903	-118.7558	25c	Small	0.0932
38 Standard and Emma	M056018	48.6306	-118.7461	25a	No production	0.0932
39 Surprise	SP00114, M056019	48.6608	-118.7492	25c	Small	0.0932
40 Tom Thumb	SP00117, M056047	48.6961	-118.7572	25c	Small	0.0932
41 Trade Dollar	M056048, SP00118	48.6733	-118.7508	25c	No production	0.0932
42 V Fraction	M056021	48.6658	-118.7458	25a	No production	0.0932
43 Zalla M	M056126, SP00122	48.7667	-118.8314	25b	Small	0.0932
44 American Flag	SP00186	48.7536	-118.8453	25b	No production	0.0932
45 Bodie Mountain Mine (Northern Gold)	M056791	48.8158	-118.9036	25a	Small	0.0932
46 Sheridan (Phil Sheridan)	SP00283	48.7786	-118.8547	25a	Small	0.0932
47 Silver Bell	SP00285	48.7597	-118.8383	25b	Small	0.0932
48 First Thought Mine	M060170, M056476	48.8839	-118.1611	25c	Small	0.0932
49 Hidden Treasure	M056495	48.8739	-118.1656	25a	No production	0.0932
50 Nest Egg	M056501	48.8481	-118.1586	25a	No production	0.0932

¹ Hot spring Au-Ag (25a); Creede epithermal veins (25b); Comstock epithermal veins (25c) ; ² Long, DeYoung, and Ludington, 1998; (.) - alternate name; na - none available

B. Placer sites

[na none available; ¹ located in Washington except those in Boundary County]

Site name	MRDS Identifier number	MAS Identifier number	Latitude	Longitude	County ¹
1 Boulder Creek placers	W007555		48.6006	-116.0883	Boundary
2 Copper Creek	na	0160210133	48.9803	-116.1728	Boundary
3 Point Bar Placer	na	0160210134	48.7861	-116.1533	Boundary
4 Alva Stout	M056072		48.6494	-118.7744	Ferry
5 Blance	M056073		48.6528	-118.8058	Ferry
6 Blue Bar Island	M056062		48.2000	-118.2003	Ferry
7 Blue Bar Island Placer	na	0530190123	48.2072	-118.1994	Ferry
8 Bridge Creek	M056063		48.2264	-118.1767	Ferry
9 Daisy	M056067		48.3897	-118.1981	Ferry
10 Dora B Placer	na	0530190127	48.6472	-118.8153	Ferry
11 Dova B	M056074		48.6528	-118.8058	Ferry
12 Gold Creek	M056077		48.4253	-118.8450	Ferry
13 Goosmus Creek Placer	na	0530190129	48.9631	-118.5806	Ferry
14 Johnson Placer	M056068		48.2911	-118.1633	Ferry
15 Keller Placers	na	0530190397	48.0856	-118.6903	Ferry
16 Ninemile	M056070		48.0158	-118.3972	Ferry
17 Sanpoint River, West Fla.	M056076		48.4581	-118.7719	Ferry
18 Singer Placer	SP00111		48.9961	-118.5308	Ferry
19 Stray Dog Placer	M056065		48.2053	-118.1992	Ferry
20 Thompson	M056066		48.2189	-118.1861	Ferry
21 Turtle Rapids Placer	na	0530190177	48.1783	-118.2500	Ferry
22 Ballard Placer	na	0530470493	48.5319	-119.7450	Okanogan
23 Cassimer Bar Placer	SP00204		48.0992	-119.7181	Okanogan
24 Condon Bar Placer	na	0530470512	48.1100	-119.3111	Okanogan
25 Crounce Placer	SP00211		48.3997	-118.8789	Okanogan
26 Cuba Line Placer	M056821		48.9967	-119.1053	Okanogan
27 Dan Mooney Placer	na	0530470513	48.9606	-119.0406	Okanogan
28 Davey Placer	na	0530470514	48.9600	-119.0408	Okanogan
29 Deadman Creek Placer	na	0530470497	48.9167	-119.0550	Okanogan
30 Fourth of July Creek Placer	na	0530470498	48.9167	-119.0558	Okanogan
31 Gold Bar Placer	na	0530470508	48.1053	-119.2664	Okanogan
32 Hopkins Placer	na	0530470499	48.1133	-119.2311	Okanogan
33 Mary Ann Creek Placer	SP00250		48.9403	-119.0517	Okanogan
34 Meadows Placer	na	0530470501	48.6550	-119.8528	Okanogan
35 Murray Placer	M056823		48.2000	-119.2544	Okanogan
36 Nespelem Bar Placer	na	0530470502	48.1361	-119.0539	Okanogan

B. Placer sites (cont.)

37 Nugget Placer	na	0530470503	48.4100	-118.9039	Okanogan
38 Rich Bar Placer	na	0530470045	48.9822	-119.5364	Okanogan
39 Shotwell Placer	SP00284		48.0342	-119.6828	Okanogan
40 Similkameen Falls Placer	na	0530470505	48.9703	-119.5017	Okanogan
41 Similkameen Placers	na	0530470506	48.9806	-119.5492	Okanogan
42 Walker Placer	SP00299		48.9689	-119.1144	Okanogan
43 Browns Lake Placer	M025803		48.4392	-117.1828	Pend Oreille
44 Harvey Bar Placer	M025804		48.9439	-117.3300	Pend Oreille
45 Schierding Placer	M025805		48.9881	-117.3447	Pend Oreille
46 Schultz Placer	na	0530510304	48.8597	-117.4111	Pend Oreille
47 Sullivan Creek Placer	na	0530510305	48.8383	-117.2653	Pend Oreille
48 Ambrose Mining	na	0530650884	48.9611	-117.8500	Stevens
49 Blue Bar	M056523		48.1714	-118.1875	Stevens
50 Blue Bar Placer	na	0530650172	48.1700	-118.1833	Stevens
51 Bossburg Placer	M056520		48.7564	-118.0475	Stevens
52 China Bend Placer	na	0530650173	48.8006	-118.0192	Stevens
53 Collins	M056525		48.3883	-118.1697	Stevens
54 Evans Placer	na	0530650174	48.9350	-117.7631	Stevens
55 Gibson Bara	M056524		48.0214	-118.3919	Stevens
56 Holsten	M056522		48.5000	-118.1750	Stevens
57 Marcus Placer	M056518		48.6656	-118.0675	Stevens
58 Meyers Falls	M056515		48.5936	-118.0647	Stevens
59 Nigger Creek Bar Placer	na	0530650179	48.9403	-117.7656	Stevens
60 Ninemile Bar	M056519		48.7956	-118.0036	Stevens
61 Nobles Placer	na	0530650182	48.8472	-117.9117	Stevens
62 Orient	M056521		48.8611	-118.1997	Stevens
63 Orient Placer	na	0530650184	48.8606	-118.0722	Stevens
64 Reed and Roberts Placer	na	0530650185	48.9544	-117.7344	Stevens
65 Sandoz	M056517		48.6964	-118.0164	Stevens
66 Stranger Creek	M056069		48.3083	-118.1478	Stevens
67 Valbush Bar	M056516		48.7000	-118.0206	Stevens

C. Table 2b—Pattern analysis results for lithologic map units

[lithology units defined in Boleneus and Causey, 2000; (1) Outside – outside the pattern]

Continued from bottom of previous column

Formation symbol	Area, km ²	Number of training sites	Rank(1)	Formation symbol	Area, km ²	Number of training sites	Rank (1)
bx	6.4	0	Outside	Ols	117.3	0	Outside
Ccbl	1.8	0	Outside	Omd	109.3	0	Outside
Ccbm	22.0	0	Outside	Oml	195.1	0	Outside
CDcb	2.8	0	Outside	pChm	166.7	0	Outside
CDmm	110.2	0	Outside	pJmm	3.6	0	Outside
CDmv	4.8	0	Outside	pJmsg	62.6	0	Outside
COcb	2.3	0	Outside	pJmx	99.9	0	Outside
COcg	6.7	0	Outside	pJtz	1.8	0	Outside
COM	62.7	0	Outside	pKma	22.4	0	Outside
COMv	29.3	0	Outside	pKmog	54.7	0	Outside
Cphm	327.9	0	Outside	pKmu	21.5	0	Outside
Czq	364.5	0	Outside	pKmx	0.9	0	Outside
Dcb	0.5	0	Outside	PLMcg	9.3	0	Outside
Ecg	27.6	0	Outside	PMmc	32.2	0	Outside
Ei	1.6	0	Outside	PMms	300.4	0	Outside
Eia	13.5	0	Outside	PMmv	2.3	0	Outside
Eib	9.3	0	Outside	pTma	31.1	0	Outside
Eig	74.8	0	Outside	pTmbg	377.2	0	Outside
Eigd	43.2	0	Outside	pTmgg	172.7	0	Outside
Eik	9.8	0	Outside	pTmi	4.2	0	Outside
Eim	50.3	0	Outside	pTmn	38.0	0	Outside
Eimd	198.0	0	Outside	pTmog	584.9	0	Outside
Eiqm	338.7	0	Outside	pTmpg	103.8	0	Outside
Eir	0.4	0	Outside	pTmq	366.6	0	Outside
EPia	6.2	0	Outside	pTmx	40.5	0	Outside
EPid	3.1	0	Outside	PZmc	4.3	0	Outside
EPig	198.6	0	Outside	PZmd	2.1	0	Outside
EPigb	330.5	0	Outside	PZmq	15.4	0	Outside
EPigd	162.0	0	Outside	PZms	36.1	0	Outside
EPigk	610.0	0	Outside	PZmu	10.6	0	Outside
EPigm	330.5	0	Outside	Scg	2.6	0	Outside
EPiqp	7.3	0	Outside	Smm	2.2	0	Outside
Et	45.8	0	Outside	TKia	23.0	0	Outside
Etz	9.4	0	Outside	TKiaa	9.8	0	Outside
Ev	2.4	0	Outside	TKigd	2.8	0	Outside
Evcg	2.5	0	Outside	TKik	2.5	0	Outside
Evcl	91.2	0	Outside	TRmc	55.0	0	Outside
Evf	217.6	0	Outside	TRmig	17.2	0	Outside
Evs	10.6	0	Outside	TRms	73.0	0	Outside
Jib	16.5	0	Outside	TRmu	1.3	0	Outside
Jik	6.1	0	Outside	TRmv	5.7	0	Outside
Jiqm	75.9	0	Outside	TRPMmb	23.5	0	Outside

C. Table 2b—Pattern analysis results for lithologic map units (cont.)

Jmc	14.1	0	Outside	TRPMmd	0.9	0	Outside
Jmig	36.1	0	Outside	TRPMms	41.4	0	Outside
Jmv	136.2	0	Outside	TRPMmv	12.4	0	Outside
JTigd	192.6	0	Outside	TRPMu	5.1	0	Outside
JTiqd	177.3	0	Outside	water	425.5	0	Outside
Ju	0.0	0	Outside	Yart	7.7	0	Outside
Kcg	9.0	0	Outside	Ybbs	55.6	0	Outside
Kid	10.7	0	Outside	Ybha	23.4	0	Outside
Kig	226.6	0	Outside	Ybhq	17.1	0	Outside
Kigd	1288.9	0	Outside	Ybi	11.3	0	Outside
Kihgd	38.1	0	Outside	Ybms	3.3	0	Outside
Kim	221.3	0	Outside	Ybps	222.5	0	Outside
Kiqm	186.1	0	Outside	Ybrq	78.1	0	Outside
KJid	17.1	0	Outside	Ybsrs	37.7	0	Outside
KJigb	2.1	0	Outside	Ybss	23.6	0	Outside
KJigd	147.6	0	Outside	Ybwq	57.2	0	Outside
KJik	3.0	0	Outside	Ybwua	29.6	0	Outside
KJmgg	45.3	0	Outside	Yed	29.7	0	Outside
KJmig	392.2	0	Outside	Ymcs	40.5	0	Outside
KJmix	262.9	0	Outside	Yprc	16.6	0	Outside
KJmm	18.7	0	Outside	Ypres	21.1	0	Outside
KJmo	136.6	0	Outside	Yprl	11.6	0	Outside
KJmog	55.0	0	Outside	Yprq	9.5	0	Outside
KJmqd	5.2	0	Outside	Ypru	18.5	0	Outside
Mc	5.6	0	Outside	Ysd	16.6	0	Outside
MDcb	4.4	0	Outside	Ytar	98.3	0	Outside
Mvg	158.5	0	Outside	Zhcg	15.3	0	Outside
Mvw	158.8	0	Outside	Zhmv	67.6	0	Outside
MZia	3.7	0	Outside	Zi	5.1	0	Outside
MZid	2.7	0	Outside	Zlmv	63.5	0	Outside
MZmg	1.7	0	Outside	Zmlv	75.7	0	Outside
MZmgg	33.1	0	Outside	Zmmm	3.3	0	Outside
MZmqd	26.9	0	Outside	Zscg	97.1	0	Outside
Mzu	5.8	0	Outside	Zsl	0.8	0	Outside
Occ	66.0	0	Outside	Zsp	28.2	0	Outside
Ocs	410.3	0	Outside	Zsq	2.7	0	Outside
Ocv	40.7	0	Outside	Ztq	47.3	0	Outside
Oig	1.3	0	Outside	Zu	2.7	0	Outside

Continued on right column at top of table

Appendix II. Glossary⁶

Buffer—A polygon enclosing an area within specified distance from a point, line or polygon. In ArcView/Weights-of-Evidence, buffering is performed using Spatial Analyst so the output is always a grid (raster). The buffering function generates one or more buffers of equal distance from the input features. Input can be either vector or raster data.

Categorical weights calculation (analysis)—Refers to weights calculated for each class in an evidential theme. In ArcView-Weights-of-Evidence, categorical analysis describes one of the tables of weights that can be created using the “Calculate Theme Weights” function, distinguishing it from “Cumulative Weights”. Categorical refers to measurements made or labels given at the nominal scale of measurement. Nominal measurements are simply numerical measurements without quantitative context. Numbers assigned arbitrarily to rock types are a common geological example (Bonham-Carter, 1994, p. 41)

Conditional independence—Conditional independence of evidential themes with respect to the training points is assumed for the weights of evidence. The product of area and posterior probability summed over each unique condition is the number of points predicted by the model. A ratio is calculated by dividing the actual number of training points input to the model by this predicted number of points. The ratio will be between 1 and 0. A value of 1 (never occurs in practice) indicates conditional independence among the evidential themes used in the model. Values less than 1 indicates a conditional independence problem although the values >0.5 may produce reasonable results. See Bonham-Carter (1994) for rigorous discussion of conditional independence.

Contrast—Difference between weights, W^+ and W^- . Difference between the natural logs of conditional odds that A and B occur together and the natural log of the conditional odds that A and B do not occur together. $C = \ln(Odds \{B|A^+\}) - \ln(Odds \{B|A^-\})$; where A = evidence layer; B = training set. A rule-of-thumb for interpreting contrast values for predictor themes is given below:

If contrast value is:	Level of prediction is:
0-0.5	Mildly
0.5-1	Moderately
1-2	Strongly
>2	Extremely

The contrast values of 1 and 2, respectively, approximate probability values of 0.75 and 0.88. The level of significance of contrast values is determined by the studentized contrast value. See Bonham-Carter (1994, p. 323) for discussion. This is the contrast divided by its standard deviation. The approach used here is that a studentized contrast value of 2.0 is approximately equivalent to a 98% level of confidence.

The relationship of probability, odds and weight (natural logarithm of odds) are shown in the table (below).

<i>Probability (P)</i>	<i>Odds</i>	<i>Weight⁽¹⁾</i>
0.1	1/9	-2.2
0.5	1/1 (even)	0.0
0.75	3/1	1.1
0.88	88/12	2.0
0.99	99/1	4.6

⁽¹⁾ also known as logit

⁶ From user guide (Kemp and others, 1999). For additional explanation, see also Bonham-Carter (1994).

Cumulative weights (analysis)—Refers to weights calculated for cumulative number of points and areas for classes of ordered data. Cumulative weights calculated from either highest to lowest (descending) or lowest to highest (ascending) class, can be calculated for a single evidential theme in the “Calculate Theme Weights” function. Refers to a method of calculating weights for cumulative distances, and examining the weights and contrasts at successive cumulative distance intervals from a source (line, point, or polygon). Calculating cumulative weights can be useful in reducing noise from variation that occurs in categorical weights, making it easier to determine the optimum cut-off points for generalization of data.

Evidence (predictor) theme—A spatial data set used as evidence for prediction of training points (e.g. mineral occurrences). ArcView-WofE is able to use polygon features themes (shapefiles or coverages) or integer grid themes (grid format) as evidential themes.

Pattern generalization—In ArcView/WofE, the product resulting from reclassification of the thematic information of an evidential theme by classifying (grouping) existing classes in the theme attribute table to fewer classes in a new field.

Negative weight, W^- —Natural logarithm of the quantity: Odds that the evidence layer and training set do not occur together divided by the odds of training set occurring within the study area. $W^- = \ln(\text{Odds}\{B|A^-\}/\text{Odds}\{B\})$; where A = evidence layer; B = training set

Normalized contrast—Contrast divided by the standard deviation of contrast.

Positive weight, W^+ —Natural logarithm of the quantity: Odds that the evidence layer and training set occur together divided by the odds of training set occurring within the study area. Difference between the unconditional or prior logit of A and the posterior logit of A. A logit equals the ln odds. $W^+ = \ln(\text{Odds}\{B|A^+\}/\text{Odds}\{B\})$; where A = evidence layer; B = training set.

Posterior probability—A redistribution of the prior probability based on the weights. See Bonham-Carter (1994) for a rigorous discussion.

Prior probability—Number of points in training set divided by the study area, expressed by the same area unit (cell size).

Response theme—An output map that expresses the probability that a unit area contains a training point, estimated by combining the weights of the predictor variable (evidence themes). The theme is based on a unique conditions grid and its attribute table.

Training set or sites—Point feature theme used in the calculation of weights. The set of spatial objects whose locations are to be predicted. In mineral exploration, these are the sites of known mineral deposits. Points are either present or absent. Size or other attributes of these points are not modeled (Raines, Bonham-Carter, and Kemp, 2000).

Weights calculation example—Weights calculations are carried out by two different procedures, by categorical or by cumulative calculations. The categorical method is used where data occurring that occur in categories being measured at not related or categories are mutually exclusive. When number or symbols are used to identify the groups to which various objects belong, the numbers or symbols are referred to as belonging to a nominal or classificatory scale of measurement (Siegel, 1956, p. 22). The categories *Yes* and *No* are an example. The lithologic units of a geologic map are another example. The *positive weight* equals the natural logarithm of the odds that the evidence layer and the training set *occur together* divided by the odds of the training set occurring in the study area. The *negative weight* equals the natural logarithm of the odds that the evidence layer and the training set *do not occur together* divided by the odds of the training set occurring within the study area. The data needed to do the calculation is gathered in a

two, 2-by-n contingency tables (where n=number of lithologic units on the map) for training sites that occur inside each lithologic unit and training sites that occur outside each lithologic unit. The Eck unit in the Klondike Mountain Formation is the example used in the table below (areas are km²):

Lithologic unit	Occur inside lithologic unit		Occur outside lithologic unit		Total	
	Training sites	Area, km ²	Training sites	Area, km ²	Training sites	Area, km ²
Eck	4	26.3	46	23,573.7	50	23,600
and so on						

Take for example, the geologic map unit Eck, the positive weight equals \ln of the number of training sites divided by the area of Eck divided by the prior probability. The prior probability is 50/23600 (the study area is 23600 km² and a training site is assumed to occupy one km²) equals .00212.

$$\ln([4/26.3]/0.00212) = 4.273$$

For Eck, the negative weight equals the \ln of the number of training sites outside Eck divided by the area outside Eck.

$$\ln([50-4/23600-26.3]/0.00212) = -.08292$$

The contrast equals the difference between the positive and negative weights

$$4.273 - (-0.08292) = 4.359$$

The cumulative weights calculation is carried out in the same fashion as outlined above. The only exception is the manner of collecting the data for the 2-by-n contingency table. Unlike the data in unrelated categories discussed above, where data are related between categories, the calculation may be carried out on a cumulative weights fashion. In this case a series of buffer bands (buffer widths vary from 100m to 1000 widths) are constructed around the features. Using 1000m wide bands, calculations are carried out for each band, 1000m, 2000m and so on, in a cumulative fashion or beyond the distance needed to capture all training sites within the buffer bands. Calculations outlined above are the same except that the area within the bands and the training sites within them are accumulated for each successive calculation (e.g. 0-1000m, 0-2000m, 0-3000m ...). Example data for a contingency table of sites occurring inside the bands is given below:

Buffer	Cumulative	Training sites	Cumulative	Area, km ²	Cumulative
1000m	1-1000m	1	1	70	70
2000	0-2000	10	11	80	160
3000	0-3000	18	29	1000	260
and so on					