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# Microscale distribution patterns of terrestrial bryophytes in a subalpine forest: the use of logistic regression as an interpretive tool

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**Abstract:** This study investigated microhabitat relationships of terrestrial bryophytes in a subalpine forest of coastal British Columbia. Substratum affinities were characterized for dominant bryophytes. Logistic regression analysis was used to gain insight into the ecological determinants of fine scale  $(0.1m^2)$  bryophyte distribution by examining the predictive relationship between bryophyte species occurrence and localized environmental conditions, as well as the coverage of other bryophytes. The predictive relationships were compared to evaluate the relative importance of environmental factors versus interspecific interactions in structuring bryophyte communities. The results indicate that bryophytes show unique responses in their relationships to environmental conditions and other bryophytes. Positive feedback appears to be an important process among terrestrial bryophytes in subalpine forests.

Nomenclature: Lawton (1971) for mosses; Godfrey (1977) for liverworts; Hitchcock and Cronquist (1973) for vascular plants.

Abbreviations: CRS - Creeping Stem; EXH - Exposed Humus; FLI - Fine Litter; LFH - The accumulation of organic material present over mineral soil (Litter, Fermentation and Humus); LR - Logistic regression; WOOD - Woody Debris.

## Introduction

Bryophyte distribution patterns on a fine scale are likely to be strongly influenced by microscale variation in substratum, neighbouring bryophytes, and other small scale habitat features (Carleton 1990). Previous studies have demonstrated the sensitivity of bryophytes to changes in microhabitat conditions (Kenkel and Bradfield 1986, Soderstrom 1993, Rkland 1994, McAlister 1995, Peck et al. 1995, Rambo and Muir 1998). Limited microscale ecological information exists for bryophytes in the coniferous forests of Pacific North America, and most of this is contained in floristic works (e.g., Lawton 1971, Schofield 1976, Godfrey 1977) and field guides (Vitt et al. 1988, Schofield 1992). Subalpine forests have been particularly underrepresented in bryophyte research. As subalpine forests face increasing pressures from recreational use and logging, more research is required to sustain high bryophyte species diversity in these fragile ecosystems.

A limitation in describing microscale bryophyte-environment relationships is that most statistical techniques inadequately convey information at the species scale. Ordination techniques, most commonly used to identify environmental gradients correlated with compositional change in plant communities (Gauch 1982), are limited in this way. As an alternative to ordination, logistic regression (LR) may be used when interest focuses directly on the responses of particular species to environmental factors. Because logistic regression analysis employs mixtures of continuous and discrete variables (e.g., microhabitat variables) to predict a discrete outcome in a response variable (e.g., presence or absence of a given species), more detailed ecological questions can be addressed than by using standard regression and ordination methods (Trexler and Travis 1993, Tabachnick and Fidell 1996).

In addition to answering specific questions regarding microscale distributions, LR may be a more statistically appropriate technique to use when data do not meet the assumptions of other approaches. For example, whereas ordination methods may assume linearity (principal component analysis), or unimodality (correspondence analysis) in the responses of species to underlying gradients, LR makes no such assumptions and requires only that response variables have minimum frequency (usually >5%) in relation to categories of predictor variables (Tabachnick and Fidell 1996). Owing to its lack of reliance on statistical assumptions, LR could find wide application in ecology (Trexler and Travis 1993). To date, however, few applications of LR have appeared in the plant ecological literature (e.g., van de Rijit et al. 1996).

The objective of this study was to describe the microscale habitat relationships of terrestrial bryophytes in a subalpine forest of coastal British Columbia. Logistic regression was used to investigate the predictability of occurrence of bryophyte species in particular environmental conditions, and in associations with other bryophytes. The results provide insight into the ecological determinants of bryophyte distribution on a fine scale, and allow for the evaluation of LR as a technique for interpreting such relationships.

#### Methods

#### Study area and field sampling

The study area is located in unlogged, old-growth subalpine forest in the Coast Mountains near Vancouver, British Columbia. The prevailing climate is characterized by cool short summers, and long, cold, wet winters with deep snow accumulation over unfrozen ground for up to ten months each year (Pojar et al. 1991). The forest canopy is comprised of *Tsuga mertensiana*, *Abies amabilis*, and *Chamaecyparis nootkatensis*, and prominent understory shrubs include *Vaccinium alaskaense*, *V. membranaceum*, *Menziesia ferruginea*, *Rhododendron albiflorum*, and *Rubus pedatus*. The terrestrial bryophyte layer is generally well-developed and comprises several species of mosses and liverworts. The mosses *Rhytidiopsis robusta* and *Dicranum pallidisetum* are common in many areas.

Forty-five 20m x 20m plots, ranging in elevation from 1100m to 1300m a.s.l., were systematically established at 50m intervals along a series of transect lines in undisturbed forest. Within each plot, eight  $0.1m^2$  microplots were systematically sampled along perpendicular transects intersecting at the plot center. This sampling design creates a data bias toward the plot center; however, it allowed optimum placement of microplots for examining contour related effects on bryophyte distributions. In total, 360 microplots were sampled.

Environmental measurements recorded at the plot scale (400m<sup>2</sup>) included landscape features (slope, aspect), and soil characteristics determined from a soil pit excavated near the plot center. Depth of the litter layer (LFH) was recorded in the field. Subsequent laboratory analysis of the soil samples yielded plot values for pH, organic matter content (loss on ignition method), and total Kjeldahl nitrogen (Carter 1993). For the purpose of data analysis the estimates of carbon and nitrogen were combined into a C:N ratio for each plot.

Environmental measurements at the microplot scale included canopy cover (measured with a spherical densiometer), and the total percent coverage and maximum height of understory vascular plants, irrespective of species. As well, the total numbers of vascular plant species and conifer seedlings (by species) in microplots were noted. The environmental conditions measured at the plot scale appeared to remain relatively constant within plots; therefore, these measurements were applied to all microplots within each plot.

Total bryophyte cover within microplots was estimated, as was the coverage of separate bryophyte species on each of four substratum-types: fine litter (FLI, consisting of conifer needles and woody debris <1cm in width), fallen branches and tree stems (WOOD, comprising woody debris 1-10cm in width), exposed humus (EXH, including decayed, compacted organic material with no buildup of fine litter or woody debris), and creeping stems (CRS, consisting of stems and branches of vascular plants lying close (<5cm) to the ground surface). Also recorded was the total cover (i.e., availability) of each of the four substratum-types within microplots.

Two features of the sampling design used warrant some concern: first, the treatment of microplots as replicates in this analysis may have introduced problems with spatial autocorrelation, particularly since bryophyte distributions are often patchy; second, the inference that some plot variables (i.e., aspect, slope, LFH characteristics) are constant at the microplot level may not be valid in all situations. In light of these concerns, the precise levels of significance associated with the LR results should be viewed with caution.

#### Data analysis

Logistic regression models produced large parameter estimates and standard errors, or failed to converge, for species with less than 5% occurrence in microplots. In consequence, only those bryophytes with greater than 5% frequency were included in further analyses. Of the 42 species of mosses and liverworts identified during sampling, 14 species met this requirement. To facilitate interpretation of bryophyte-substratum affinities, the frequencies of these 14 bryophytes on the four main sub- stratumtypes were calculated as the number of microplots in which a species occurred on a particular substratum divided by the total number of microplots containing that substratum.

To investigate the relationship between bryophyte species occurrences and microplot conditions, two types of logistic regression were performed on each species: one using the environmental variables (landscape, soil, vascular plant cover and species richness, substratum availability) as predictors, and one using the coverages of the other 13 bryophyte species as predictors.

Although LR is free of parametric restrictions, and can analyze a mixture of types of predictors (continuous, discrete, and dichotomous), it is required that the outcome variable be discrete (in this case, the presence or absence of bryophytes). Also, the method requires an adequate ratio of cases to variables; it is recommended that  $N \ge 50 +$ 80m (where N = sample size, and m = number of independent variables) for testing regression, and  $N \ge 104 + m$ for testing individual predictors (Tabachnick and Fidell 1996). These recommendations were met for the LR analyses in this study. Moreover, LR is sensitive to high correlations among predictor variables (multicollinearity), and can be influenced by outliers. Fine litter (FLI) was not used as a predictor variable in the environmental LRs because of its strong negative correlation with the availability of woody debris (WOOD).

The goal of logistic regression is to find the best linear combination of predictors to maximize the likelihood of obtaining the observed outcome frequencies. The linear regression equation is the natural log of the probability of being in one group divided by the probability of being in the other group. LR predictions take the form of a probability p ranging from 0 and 1; the probabilities are calculated using maximum likelihood estimation. Classification is usually accomplished by rounding the p value to the nearest integer (either 0 or 1).

All LR analyses in this study employed the LOGIT program of SYSTAT® Version 8.0 (SPSS Inc. 1998). Models were tested in a direct (rather than stepwise) fashion. Wald's statistics were used to indicate the importance of individual microhabitat variables in predicting the presence or absence of a bryophyte species. The sign of the logit coefficients suggested whether a species was more likely (positive sign) or less likely (negative sign) to occur given a change in a microhabitat variable. Rho-squared values (analogous to R<sup>2</sup>-values in multiple regression) were used to characterize the overall strength of the relationships between bryophyte presence/absence and the set of microhabitat (environment or bryophyte) predictor variables. According to Tabachnick and Fidell (1996), rho-squared values greater than 0.2 indicate a satisfactory relationship between predictor and response variables in LR. All results were adjusted using the quasi-maximum likelihood procedure (a component of the SYSTAT® LOGIT program), which corrects covariance problems in models.

#### **Results and discussion**

## Substratum frequencies

Table 1 shows the percent frequencies of bryophytes recorded on the four main substratum-types. In general, mosses occur with higher frequencies than liverworts on fine litter (FLI). Liverwort species occur mainly on exposed humus (EXH), a prominent substratum-type in closed-canopy habitats that have a build-up of undecomposed organic material. *Pseudoleskea baileyi* is the only species that occurs primarily on the creeping stems of vascular plants. The mosses *Hypnum circinale* and *Rhyti-diopsis robusta*, and the liverwort *Ptilidium californicum*, show relatively wide tolerance ranges, occurring on each of the four substratum-types with at least 5% frequency.

Although this information increases our understanding of the occurrence of bryophyte species in subalpine forests, substratum affinity is only one mechanism that may influence the presence or absence of species. For example, species occurrence may be more related to a particular set of environmental conditions than the availability of a certain substratum when that substratum is widely available in different habitats. Furthermore, the occurrence of a bryophyte species may be related to the presence, absence, or coverage of other bryophyte species.

#### Microscale prediction of bryophyte species occurrence

The predictive relationships of bryophyte species with environmental factors and coverages of other bryophytes in microplots are summarised in Tables 2 and 3, respectively. The results indicate substantial differences among species in the predictability of occurrence from environmental factors and the coverage of other bryophytes. The differing response patterns reveal habitat features which may be of interest in more detailed autecological studies. Among the environmental variables, exposed humus availability is a positive predictor for five of the eight liverwort species analyzed, suggesting that it is an important environmental factor contributing to bryophyte diversity. **Table 1**. Percent frequencies of mosses and liverworts with >5%\* microplot frequency (n is the total number of microplots in which the species occurred) on fine litter (FLI), woody debris(WOOD), exposed humus (EXH), and creeping stem (CRS) substratum-types in relation to the total number of microplots containing each substratum-type (shown at the top of each column).

			FLI	WOOD	EXH	CRS
Code	Mosses	n	(360)	(303)	(71)	(50)
Braclei	Brachythecium leibergii	25	5.8	0.7	5.6	2.0
Dicrpal	Dicranum palledisetum	254	62.2	9.6	39.4	4.0
Hypncir	Hypnum circinale	74	12.8	6.9	23.9	2.0
Plaglae	Plagiothecium laetum	94	15.0	7.9	33.8	2.0
Pseubai	Pseudoleskea baileyi	64	13.6	4.6	4.2	42.0
Rhytrob	Rhytidiopsis robusta	217	59.7	6.9	16.9	12.0
Average l	Frequency (Mosses):		28.2	6.1	20.7	10.7
Code	Liverworts					
Barbflo	Barbilophozia floerkei	33	8.3	1.0	2.8	0.0
Bleptri	Blepharostoma trichophyllum	35	2.8	3.3	25.4	0.0
Calymue	Calypogeia muelleriana	47	7.5	2.6	22.5	0.0
Cephlun	Cephalozia lunulifolia	49	4.4	5.0	31.0	2.0
Dipltax	Diplophyllum taxifolium	18	3.9	0.3	5.6	0.0
Lophhet	Lophocolea heterophylla	100	13.9	9.6	42.3	4.0
Lophgut	Lophozia guttulata	35	5.8	3.6	11.3	2.0
Ptilcal	Ptilidium californicum	125	13.3	22.8	19.7	28.0
Average l	Frequency (Liverworts):		8.3	8.3	22.0	7.2

Bryophytes with <5% frequency in microplots included 20 mosses: Antitrichia curtipendula, Dicranella heteromalla, Digiotrichum parallelum, Plagiothecium undulatum, Pleurozium schreberi, Pohlia nutans, Polytrichastrum alpinum, Pseudoleskea stenophylla, Pseudotaxiphyllum elegans, Pterigynandrum filiforme, Racomitrium ericoides, Racomitrium sudeticum, Rhizomnium nudum, Rhytidiadelphus loreus, Rhytidiadelphus squarrosus; and 8 liverworts: Cephalozia bicuspidata, Diplophyllum plicatum, Geocalyx graveolens, Lepidozia reptans, Porella cordaeana, Scapania bolanderi, Scapania diplophylloides, Scapania scandica.



**Figure 1.** Summary of logistic regression models relating presence/absence of mosses (open circles) and liverworts (solid circles) to environmental variables (ENVIRO) and coverage of other bryophyte species (BRYO). Axes are Rho-squared values denoting overall strength of the predictive relationships.

**Table 2**. Probabilities (p-values) from logistic regression models predicting the presence (+) or absence (-) of each bryophyte species in microplots from environmental variables<sup>\*</sup>. Significant p-values ( $\leq 0.05$ ) are shown in bold.

Predictor	Bra	aclei	Dic	rpal	Hy	oncir	Pla	glae	Pse	ubai	Rhy	trob	Bai	bflo
variables	р	effect	р	effect										
C:N Ratio	0.01	-	0.17	+	0.00	-	0.04	-	0.75	-	0.17	+	0.30	+
Avg pH	0.00	-	0.01	-	0.01	-	0.75	+	0.01	-	0.22	-	0.92	+
LFH	0.07	-	0.24	-	0.05	+	0.63	+	0.02	-	0.08	-	0.03	-
Slope	0.05	+	0.83	-	0.85	-	0.62	+	0.58	-	0.15	-	0.25	-
Aspect	0.72	+	0.17	-	0.71	-	0.39	-	0.00	+	0.27	-	0.01	-
AvgCan	0.21	-	0.01	-	0.30	+	0.43	+	0.12	-	0.42	+	0.03	-
AvgVasc	0.89	-	0.27	-	0.59	+	0.08	-	0.83	-	0.52	-	0.04	-
HtVasc	0.24	-	0.16	+	0.74	-	0.17	+	0.05	+	0.76	+	0.53	+
NumVasc	0.15	+	0.88	-	0.00	-	0.59	+	0.11	+	0.64	+	0.14	+
NumSeed	0.43	+	0.20	+	0.38	+	0.02	+	0.44	-	0.30	-	0.53	-
%EXH	0.39	+	0.11	+	0.16	+	0.04	+	0.61	-	0.90	+	0.49	-
%CRS	0.86	+	0.86	+	0.44	-	0.27	+	0.12	+	0.15	-	0.17	-
%WOOD	0.45	+	0.09	-	0.70	+	0.28	-	0.07	-	0.00	-	0.35	-
Predictor	Ble	eptri	Cal	ymue	Сер	hlun	Dip	oltax	Lop	hhet	Lop	hgut	Ptilcal	
variables	р	effect	р	effect										
C:N Ratio	0.82	+	0.34	-	0.53	+	0.41	-	0.20	-	0.75	+	0.21	-
Avg pH	0.84	-	0.94	-	0.19	-	0.54	+	0.53	+	0.37	+	0.00	-
LFH	0.34	-	0.31	-	0.75	+	0.40	-	0.68	+	0.05	-	0.14	-
Slope	0.18	+	0.53	+	0.14	+	0.09	+	0.59	+	0.39	+	0.41	+
Aspect	0.45	-	0.02	-	0.19	-	0.34	-	0.05	+	0.08	-	0.66	+
AvgCan	0.69	-	0.21	-	0.39	+	0.03	-	0.14	+	0.00	-	0.28	+
AvgVasc	0.94	+	0.73	+	0.33	+	0.46	-	0.93	+	0.62	+	0.59	-
HtVasc	0.81	-	0.62	+	0.54	+	0.05	+	0.24	+	0.85	-	0.49	+
NumVasc	0.44	+	0.09	+	0.61	-	0.26	-	0.46	+	0.44	+	0.91	-
NumSeed	0.95	+	0.50	+	1.00	-	0.40	-	0.17	+	0.23	+	0.20	+
%EXH	0.00	+	0.00	+	0.00	+	0.79	-	0.02	+	0.85	-	0.03	+
%CRS	0.70	-	0.00	-	0.41	-	0.07	-	0.13	-	0.05	-	0.03	+
%WOOD	0.01	+	0.79	-	0.39	+	0.01	-	0.63	+	0.59	+	0.00	+

\* Variables shown include carbon to nitrogen ratio (C:N ratio); average pH (Avg pH); LFH depth in cm (LFH); slope in degrees (Slope); aspect in degrees from north (Aspect); average canopy coverage (AvgCan); average vascular plant coverage (Avgvasc); maximum vascular plant height (HtVasc); number of species of vascular plants (Numvasc); number of conifer seedlings (NumSeed); availability (i.e., coverage) of substrata (%EXH, %CRS, %WOOD).

The pattern of significance among soil variables (C:N ratio, pH, LFH) for mosses suggests that they may be more sensitive to changes in soil chemistry than are liverworts. Where significant, average canopy cover and average cover of vascular plants show negative predictions of bryophyte occurrence.

When the coverages of other bryophyte species are used as predictors, most of the significant relationships are positive (Table 3). Two exceptions are the reciprocal negative predictions between *Hypnum circinale* and *Barbilophozia floerkei*, and the negative prediction of coverage of *Calypogeia muelleriana* for *Rhytidiopsis robusta*. Both cases involve a moss-liverwort relationship, and suggest that microhabitat distinctions for some species may be pronounced.

The predominance of positive relationships between species occurrences and the coverages of other species suggests that interspecific facilitation among terrestrial bryophytes may have an important influence on distribution patterns in subalpine forests. That is, bryophyte species, once established, may facilitate the establishment of other bryophytes. This example of positive feedback among bryophytes is similar to those described for plantplant and plant-soil interactions in other ecosystems (Perry et al. 1989, Cox and Larson 1993, Rietkerk and van de Koppel 1997).

**Table 3**. Probabilities (p-values) from logistic regression models predicting the presence (+) or absence (-) of each bryophyte species in microplots from percent coverages of other bryophyte species. Significant p-values ( $\leq 0.05$ ) are shown in bold.

Predictor Braclei		Dicrpal Hypncir		Plaglae		Pseubai		Rhytrob		Barbflo					
variables	р	effect	р	effect	р	effect	р	effect	р	effect	р	effect	р	effect	
Braclei			0.26	-	0.84	+	0.92	+	0.30	-	0.38	+	0.71	-	
Dicrpal	0.02	+			0.82	-	0.26	+	0.00	+	0.73	+	0.00	+	
Hypncir	0.21	-	0.25	+			0.03	+	0.49	-	0.60	-	0.00	-	
Plaglae	0.57	-	0.10	-	0.73	+			0.52	-	0.15	-	0.21	-	
Pseubai	0.39	+	0.93	-	0.40	-	0.10	-			0.61	+	0.49	-	
Rhytrob	0.32	+	0.14	+	0.11	-	0.56	-	0.06	+			0.73	-	
Barbflo	0.21	-	0.43	+	0.00	-	0.27	-	0.06	+	0.03	+			
Bleptri	0.94	+	0.96	-	1.00	+	0.45	-	0.37	+	0.86	+	0.81	+	
Calymue	0.00	+ .	0.55	+	0.27	-	0.19	+	0.45	-	0.01	-	0.07	+	
Cephlun	0.00	+	0.15	+	0.04	+	0.47	-	0.12	-	0.30	+	0.66	+	
Dipltax	0.82	+	0.33	-	0.48	+	0.98	-	0.26	+	0.18	+	0.99	-	
Lophhet	0.42	+	0.05	+	0.27	+	0.02	+	0.75	-	0.21	+	0.81	-	
Lophgut	0.89	+	0.06	+	0.33	-	0.08	-	0.11	-	0.82	+	0.02	+	
Ptilcal	0.53	+	0.30	+	0.05	+	0.07	+	0.17	-	0.18	-	0.67	+	
															_
Predictor	Ble	ptri	Caly	mue	Сер	hlun	Dip	ltax	Lop	hhet	Lop	hgut	Pti	lcal	
Predictor variables	Ble p	<b>ptri</b> effect	Caly p	mue effect	Cep p	<b>hlun</b> effect	Dip p	ltax effect	<b>Lop</b> p	hhet effect	Lop p	hgut effect	Pti p	ilcal effect	
<b>Predictor</b> variables Braclei	<b>Ble</b> p 0.38	effect	<b>Caly</b> p 0.31	effect	Cep p 0.39	hlun effect -	<b>Dip</b> p 0.37	effect	<b>Lop</b> p 0.10	hhet effect	<b>Lop</b> p 0.44	hgut effect	Pti p 0.88	ilcal effect	
<b>Predictor</b> variables Braclei Dicrpal	Ble p 0.38 0.21	e <b>ffect</b> - +	Caly p 0.31 0.95	effect +	Cep p 0.39 0.78	hlun effect - -	Dip p 0.37 0.35	oltax effect - +	<b>Lop</b> p 0.10 0.90	hhet effect - +	Lop p 0.44 0.00	hgut effect - +	Pti p 0.88 0.53	ilcal effect - +	
<b>Predictor</b> variables Braclei Dicrpal Hypncir	Ble p 0.38 0.21 <b>0.01</b>	e <b>ptri</b> effect - + +	Caly p 0.31 0.95 0.95	y <b>mue</b> effect - + +	Cep p 0.39 0.78 0.06	hlun effect - - +	Dip p 0.37 0.35 0.08	ltax effect - + -	<b>Lop</b> p 0.10 0.90 <b>0.00</b>	hhet effect - + +	<b>Lop</b> p 0.44 <b>0.00</b> 0.48	hgut effect - + -	Pti p 0.88 0.53 0.37	ilcal effect - + +	
Predictor variables Braclei Dicrpal Hypncir Plaglae	Ble p 0.38 0.21 0.01 0.18	e <b>ptri</b> effect - + + +	Caly p 0.31 0.95 0.95 0.00	ymue effect + + +	Cep p 0.39 0.78 0.06 0.86	hlun effect - + -	Dip p 0.37 0.35 0.08 0.66	ltax effect + - +	Lop p 0.10 0.90 0.00 0.08	hhet effect + + +	Lop p 0.44 0.00 0.48 0.06	hgut effect - + - +	Pti p 0.88 0.53 0.37 0.98	ilcal effect + +	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai	Ble p 0.38 0.21 0.01 0.18 0.35	effect - + + +	Caly p 0.31 0.95 0.95 0.00 0.60	y <b>mue</b> effect + + +	Cep p 0.39 0.78 0.06 0.86 0.67	hlun effect - - + -	Dip p 0.37 0.35 0.08 0.66 0.65	ltax effect + - + -	Lop p 0.10 0.90 0.00 0.08 0.35	hhet effect + + +	Lop p 0.44 0.00 0.48 0.06 0.46	hgut effect + - + -	Pti p 0.88 0.53 0.37 0.98 0.32	lcal effect + + -	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai Rhytrob	<b>Ble</b> p 0.38 0.21 <b>0.01</b> 0.18 0.35 0.14	effect - + + + +	Caly p 0.31 0.95 0.95 0.00 0.60 0.25	ymue effect + + + +	Cep p 0.39 0.78 0.06 0.86 0.67 0.33	hlun effect - + - - +	Dip p 0.37 0.35 0.08 0.66 0.65 0.55	ltax effect + - + - -	<b>Lop</b> p 0.10 0.90 <b>0.00</b> 0.08 0.35 0.46	hhet effect + + + -	Lop p 0.44 0.00 0.48 0.06 0.46 0.50	hgut effect + - + - + + -	Pti p 0.88 0.53 0.37 0.98 0.32 0.63	lcal effect + + - -	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai Rhytrob Barbflo	Ble p 0.38 0.21 0.01 0.18 0.35 0.14 0.01	<b>ptri</b> effect + + + + +	Caly p 0.31 0.95 0.95 0.00 0.60 0.25 0.93	y <b>mue</b> effect + + + +	Cep p 0.39 0.78 0.06 0.86 0.67 0.33 0.01	hlun effect - + - - + + + +	Dip p 0.37 0.35 0.08 0.66 0.65 0.55 0.66	ltax effect + - + - + - +	<b>Lop</b> p 0.10 0.90 <b>0.00</b> 0.08 0.35 0.46 0.14	hhet effect + + - -	Lop p 0.44 0.00 0.48 0.06 0.46 0.50 0.05	hgut effect - + - + + + + +	Pti p 0.88 0.53 0.37 0.98 0.32 0.63 0.33	lical effect + + - - +	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai Rhytrob Barbflo Bleptri	Ble p 0.38 0.21 0.01 0.18 0.35 0.14 0.01	<b>ptri</b> effect + + + + +	Caly p 0.31 0.95 0.95 0.00 0.60 0.25 0.93 0.11	y <b>mue</b> effect + + + + + + + +	Cep p 0.39 0.78 0.06 0.86 0.67 0.33 0.01 0.30	hlun effect - + - + + + + +	Dip p 0.37 0.35 0.08 0.66 0.65 0.65 0.66 0.92	ltax effect - + - + - + - + + +	<b>Lop</b> p 0.10 0.90 <b>0.08</b> 0.35 0.46 0.14 0.31	hhet effect + + + - - -	Lop p 0.44 0.00 0.48 0.06 0.46 0.46 0.50 0.05 0.00	hgut effect - + - + + + + + +	Pti p 0.88 0.53 0.37 0.98 0.32 0.63 0.33 0.17	lical effect + + - - + + + +	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai Rhytrob Barbflo Bleptri Calymue	Ble p 0.38 0.21 0.01 0.18 0.35 0.14 0.01 0.02	<b>ptri</b> effect + + + + + +	Caly p 0.31 0.95 0.95 0.00 0.60 0.25 0.93 0.11	y <b>mue</b> effect + + + + + + +	Cep p 0.39 0.78 0.06 0.86 0.67 0.33 0.01 0.30 0.39	hlun effect - + - + + + + + +	Dip p 0.37 0.35 0.08 0.66 0.65 0.55 0.66 0.92 0.00	ltax effect + + + + + + + + +	Lop p 0.10 0.90 0.00 0.08 0.35 0.46 0.14 0.31 0.16	hhet effect + + + - - - + -	Lop p 0.44 0.00 0.48 0.06 0.46 0.50 0.05 0.00 0.00	hgut effect - + - + + + + + + +	Pti p 0.88 0.53 0.37 0.98 0.32 0.63 0.33 0.17 0.19	ilcal effect + + - - + + + + +	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai Rhytrob Barbflo Bleptri Calymue Cephlun	Ble p 0.38 0.21 0.01 0.18 0.35 0.14 0.01 0.02 0.04	p <b>tri</b> effect + + + + + + + +	Caly p 0.31 0.95 0.95 0.00 0.60 0.25 0.93 0.11 0.05	ymue effect + + + + + + + +	Cep p 0.39 0.78 0.06 0.86 0.67 0.33 0.01 0.30 0.39	hlun effect - + - + + + + +	Dip p 0.37 0.35 0.08 0.66 0.65 0.55 0.66 0.92 0.00 0.14	ltax effect - + - + - + + + + +	Lop p 0.10 0.90 0.08 0.35 0.46 0.14 0.31 0.16 0.04	hhet effect + + - - - + + - + +	Lop p 0.44 0.00 0.48 0.06 0.46 0.50 0.05 0.05 0.00 0.00 0.02	hgut effect - + - + + + + + + + +	Pti p 0.88 0.53 0.37 0.98 0.32 0.63 0.33 0.17 0.19 0.05	ilcal effect + + - - + + + + + +	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai Rhytrob Barbflo Bleptri Calymue Cephlun Dipltax	Ble p 0.38 0.21 0.18 0.35 0.14 0.01 0.02 0.04 0.97	p <b>tri</b> effect + + + + + + + + +	Caly p 0.31 0.95 0.95 0.00 0.60 0.25 0.93 0.11 0.05 0.05	ymue effect - + + + + + + + + +	Cep p 0.39 0.78 0.06 0.86 0.67 0.33 0.01 0.30 0.39 0.45	hlun effect - + - + + + + + +	<b>Dip</b> p 0.37 0.35 0.08 0.66 0.65 0.55 0.66 0.92 <b>0.00</b> 0.14	ltax effect - + - + - + + + + +	Lop p 0.10 0.90 0.08 0.35 0.46 0.14 0.31 0.16 0.04 0.27	hhet effect + + - - - + + + + +	Lop p 0.44 0.00 0.48 0.06 0.46 0.50 0.05 0.00 0.00 0.02 0.05	hgut effect + + + + + + + + + + +	Pti p 0.88 0.53 0.37 0.98 0.32 0.63 0.33 0.17 0.19 0.05 0.51	lical effect + + - + + + + + + - +	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai Rhytrob Barbflo Bleptri Calymue Cephlun Dipltax Lophhet	Ble p 0.38 0.21 0.01 0.18 0.35 0.14 0.01 0.02 0.04 0.97 0.00	<b>ptri</b> effect + + + + + + + + + + +	Caly p 0.31 0.95 0.95 0.00 0.60 0.25 0.93 0.11 0.05 0.05 0.73	ymue effect - + + + + + + + + + + -	Cep p 0.39 0.78 0.06 0.86 0.67 0.33 0.01 0.30 0.39 0.45 0.01	hlun effect - + - + + + + + + +	<b>Dip</b> p 0.37 0.35 0.08 0.66 0.65 0.55 0.66 0.92 <b>0.00</b> 0.14 <b>0.03</b>	ltax effect - + - + - + + + + + +	Lop p 0.10 0.90 0.00 0.08 0.35 0.46 0.14 0.31 0.16 0.04 0.27	hhet effect + + - - - + + + + +	Lop p 0.44 0.00 0.48 0.06 0.46 0.50 0.05 0.00 0.02 0.02 0.88	hgut effect - + - + + + + + + + + + +	Pti p 0.88 0.53 0.37 0.98 0.32 0.63 0.33 0.17 0.19 0.05 0.51 0.45	lcal effect + + - + + + + + + - + +	
Predictor variables Braclei Dicrpal Hypncir Plaglae Pseubai Rhytrob Barbflo Bleptri Calymue Cephlun Dipltax Lophhet Lophgut	Ble p 0.38 0.21 0.01 0.18 0.35 0.14 0.01 0.02 0.04 0.97 0.00 0.04	<b>ptri</b> effect + + + + + + + + + + + +	Caly p 0.31 0.95 0.95 0.00 0.60 0.25 0.93 0.11 0.05 0.73 0.07	ymue effect - + + + + + + + + + + +	Cep p 0.39 0.78 0.06 0.86 0.67 0.33 0.01 0.30 0.39 0.45 0.01 0.72	hlun effect - + - + + + + + + + -	Dip p 0.37 0.35 0.08 0.66 0.65 0.55 0.66 0.92 0.00 0.14 0.03 0.05	ltax effect - + - + + + + + + + +	Lop p 0.10 0.90 0.00 0.08 0.35 0.46 0.14 0.31 0.16 0.04 0.27 0.17	hhet effect + + - - - + + + + +	Lop p 0.44 0.00 0.48 0.06 0.46 0.50 0.05 0.00 0.02 0.02 0.88	hgut effect - + - + + + + + + + +	Pti p 0.88 0.53 0.37 0.98 0.32 0.63 0.33 0.17 0.19 0.05 0.51 0.45 0.21	lcal effect + + - + + + + + - + - + -	

The results for both types of LR models can be compared using rho-squared values to elucidate which factors (environment and/or coverages of other bryophytes) most influence the occurrence of individual species. Fig. 1 illustrates the rho-squared values associated with bryophyte species occurrence in relation to the type of predictor variables used. Species that have a rho-squared value >0.2 on one or both axes are more predictably (i.e., less randomly) distributed. These species must have a narrower range of tolerance because they occur only where there is a specific set of environmental conditions, and/or where other bryophytes are established.

Bryophytes with a rho-squared value of > 0.2 on the *x*-axis, but < 0.2 on the *y*-axis, can be characterized as having strong relationships with environmental variables (e.g., particular substratum affinities or habitat requirements) that may restrict their fine scale distribution; inter-

active associations with other bryophytes are less important. Conversely, if the occurrence of a species can only be predicted from the coverages of other bryophytes, facilitation may be influencing distribution patterns. *Blepharostoma trichophyllum* is the only species that has a high rho-squared value on the *y*-axis alone. LR results (Table 3) demonstrate that the occurrence of this species is positively associated with the increased coverage of many of the other species.

Bryophytes that have a rho-squared value < 0.2 on both axes cannot be successfully predicted from environmental variables or from the coverage of other bryophytes in microplots; thus, they are more randomly distributed. These species are inferred to have a wider range of physiological tolerance because they occur over a broader range of environmental conditions and associations with other species.

#### Management implications

Of the fourteen most frequent terrestrial bryophytes in this study, five cannot be successfully predicted from environmental factors and the cover of other bryophytes. Because these species are present in a variety of ecological situations, they are less likely to be permanently lost from an area in the event of environmental disturbance, such as logging, compared to species with closer ties to particular habitat conditions. For the latter group of species, the significant variables in Table 2 provided guidelines for determining which environmental factors are most influential at the microplot scale. Because of the differences in responses among species for microhabitat conditions, it is not expected that disturbance will affect all terrestrial bryophytes in subalpine forests in the same fashion. If, as suggested, facilitation is an important process in these forests, some species may be maintained only when habitat disturbance is minimized.

It is noteworthy that twenty-eight terrestrial bryophyte species did not occur frequently enough to be included in this analysis. The infrequency of these species suggests they may also require protection from habitat disturbance. Further study, however, is required of bryophytes associated with other types of substrata (i.e., epiphytes and epixylics) to determine which species are truly rare in subalpine forests.

Once "environmentally sensitive" bryophyte species have been characterized, LR analysis can be used secondarily to generate management guidelines for sustaining high bryophyte species diversity in subalpine forests. To build a model for applied usage from measurable variables, stepwise logistic regression analysis (a technique that adds or removes variables in order of their numerical importance in the model) may be more appropriate than the direct method of analysis used in this study. Although the stepwise method may be used to generate statistically similar models with the fewest number of predictive variables, the variables chosen by this procedure do not necessarily infer ecological importance. The significance of predictors in a LR model will depend on which variables are included. In consequence, there is more than one avenue for analysis and the results are open to interpretation.

It is concluded that LR is a useful statistical technique for examining species-habitat relationships at a fine scale. In this study, LR provided information about which variables most influence the occurrences of terrestrial bryophyte species in subalpine forests. LR analyses also provided insight into the relative importance of interactive processes in microscale distribution patterning. The use of habitat variables to predict the occurrence of bryophyte species may have practical applications in the development of forest management strategies that attempt to promote bryophyte diversity, or preserve rare species.

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