

# Environmental, Spatial and Structural Components in the Composition of Mountain Forest in the Bavarian Alps

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**WINALP**  
Waldinformationssystem Nordalpen



# Background

- **WINALP was launched to deliver ecological planning tools for future management of mountain forests.**
- **Knowledge of ecological gradients is crucial in managing mountain forests.**
- **This study aims to contribute to this task by analysing the variation of species composition and its relationship to site factors .**

# Outline

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- **Study region**
- **Questions**
- **Sampling design**
- **Data Acquisition**
- **Data Analyses**
- **Results**
- **Conclusion**



## Study region

- **Mountain range in the Northern Alps (Bavaria)**
- **Extension 4,600 km<sup>2</sup>**

**53 % covered by forest**

- **Range from unnaturally structured pure coniferous forests to near-natural stands**

## Strong gradients

- **Altitude: 450 m to 2,962 m**
- **Mean temperature: -4 °C to +9 °C**
- **Annual precipitation: 1200 mm to 2800 mm**
- **Rock types: Sandstone, limestone, conglomerate,**

# Questions

- **Which main gradients underline compositional variation in mountain forests of the Bavarian Alps?**
- **How much compositional variation is explained by abiotic: soil, climate, spatial and by biotic predictors: tree layer .**



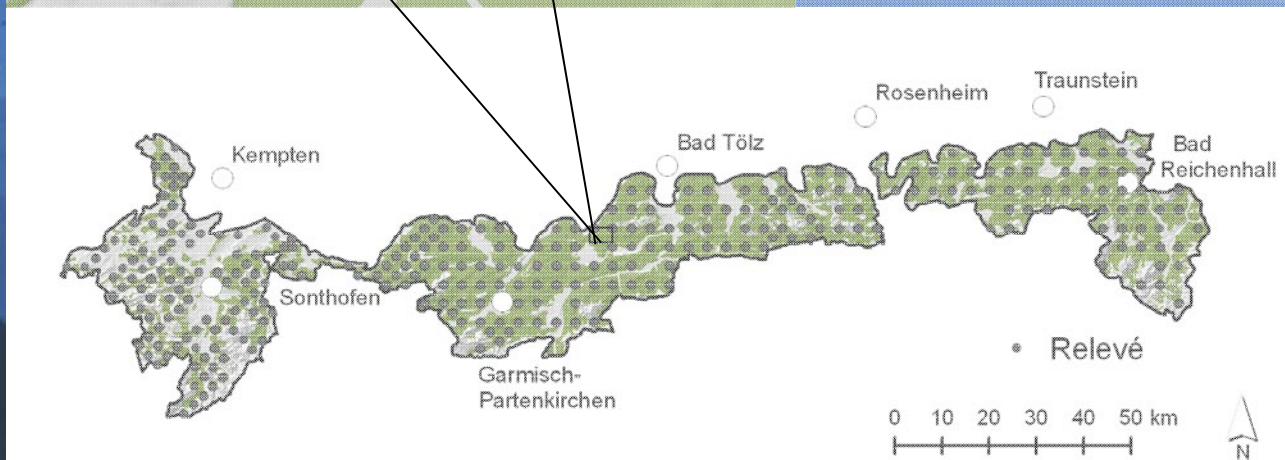
# Sampling Design

- Relevés were sampled in in a **combined systematic and stratified** design.
- The **systematic sampling** design was based on the regular 4 km x 4 km squared grid .
- **A first** inventory plot was determined by the grid of the NFI 2.
- Along a 2 km contour line forest types were continuously mapped.
- The **stratified contour line** served as the basis for the placement of **four additional** inventory plots within sections of differing forest types.

# Sampling Design



- Systematic and stratified sampling
- Grid points from NFI2 (4km)
- Mapping 2 km contour line
- Additional inventory plots within sections of differing forest types.
- 1,505 forest plots



Modified from: Reger (2011)

# Topographic Data

- **Elevation**
- **Slope**
- **Aspect**

# Vegetation Data

- **1505 plots**
- **14 m x 14 m**
- **Vascular plants and horizontal cover**
- **Six point scale: (+: <1 %, 1: 1-5 %, 2: 6-25 %, 3: >26-50 %, 4: >51-75 %, 5:> 75%)**

# Forest Types

- **Ewald J, Binner S (2007)** Werkzeuge zur Bestimmung der Waldtypen im bayerischen Hochgebirge. *Waldökologie online* (5), pp. 25–77
- **Assessment of the forest types :**
  - 47 forest types
  - Ground vegetation, stand , structure, relief and soil

# Soil data

Soil profiles were inspected in auger samples



**1505 soil profiles**

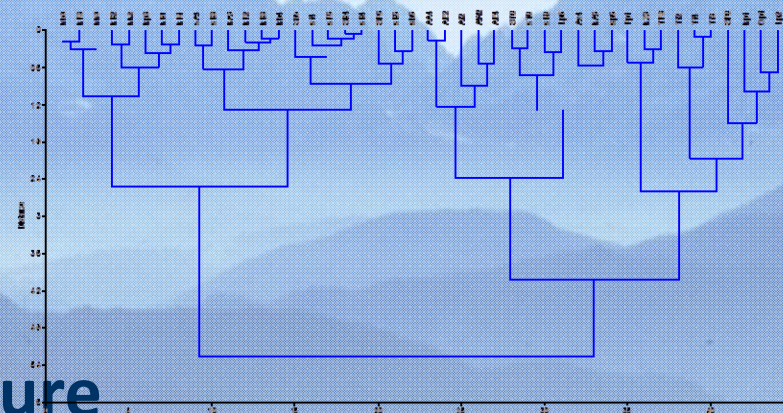
- **Morphological properties of the horizons**
- **Hydromorphic features**

# Climatic Data

- **Reger B, Kölling C, Ewald J (2011): Modelling effective thermal climate for mountain forests in the Bavarian Alps: Which is the best model? Journal of Vegetation Science 22 (4), pp. 677–687.**
- **Average monthly sums of precipitation**
- **Average monthly mean temperature**

# Available data

- 1505 releves, 910 vascular plants
- 50 forest types
- 46 site factors
  - Soil
  - Relief
  - Monthly temperature
  - Monthly precipitation





# Tree layer / Understory

- We hypothesized that tree layer composition has a strong management component due to silviculture, wildlife management and selective game browsing.
- In consequence, the analysis were performed separately for tree layer and understory.

# Tree layer / Understory

- **Tree layer influence the understory, e. g. shading, litterquality**
- **Species of the tree layers were treated as predictor variables for the understory**

# Data Preprocessing

- **Omitting species occurring only once (645 species left).**
- **Omitting species with negligible variance**
- **Finally 278 species left.**
- **Histogram transformation to dampen the influence of dominant species.**
- **Bulding overstorey (Tree1, Tree2) and understorey (Shrub, Herbs) subsets.**

# Data Analysis

**To answer the questions we applied constrained ordination (CCA) and variance partitioning.**

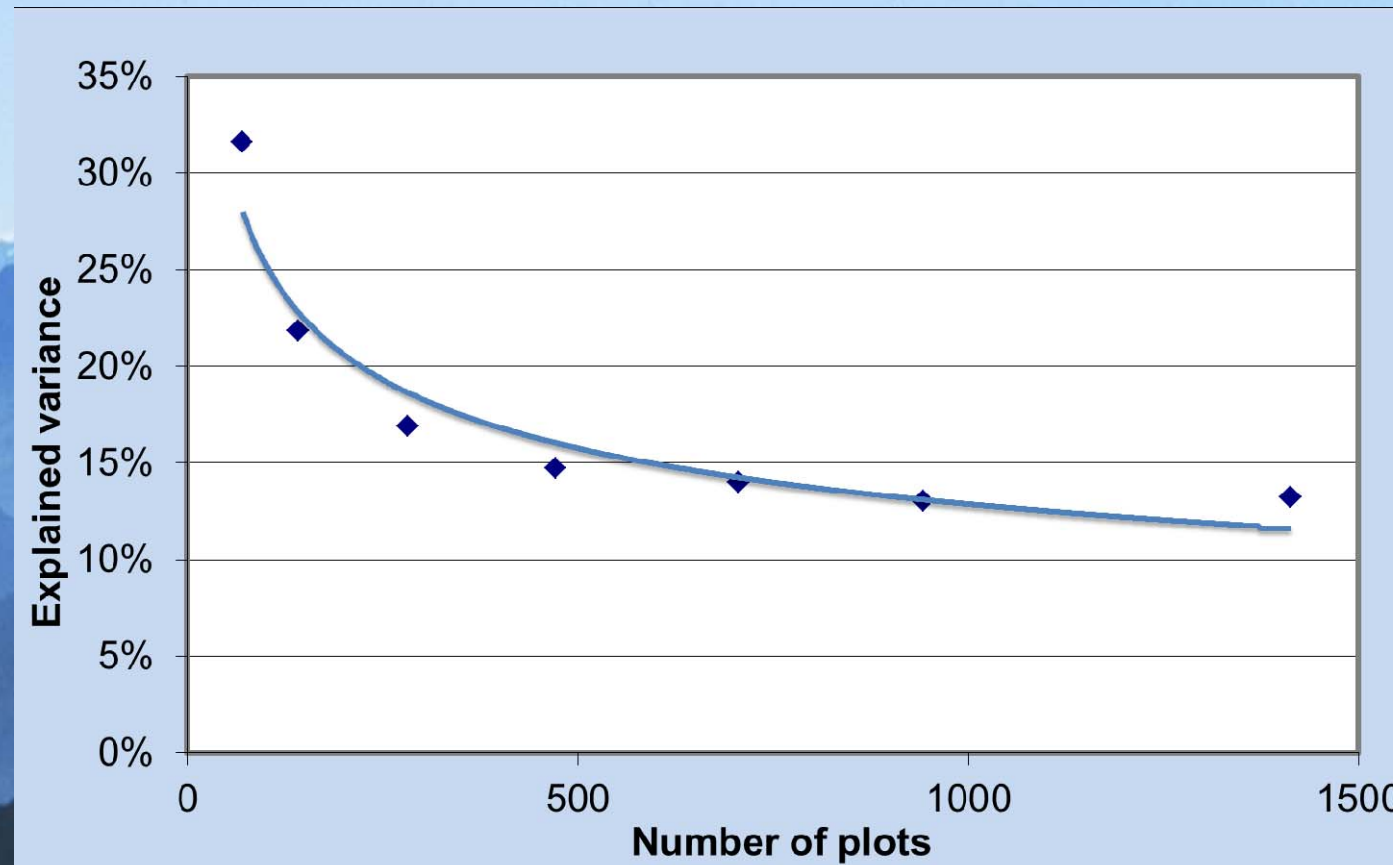
# Data Analysis

- Analyses are mainly based on canonical ordination methods using CANOCO software (ter Braak, Šmilauer 2002).
- DCA was applied to determine the length of the gradient.
- CCA was chosen as length of gradient  $\gg 2$ .
- Forward selection was applied to select significant predictor variables.

# Eigenvalues in large data sets

- At first glance explained variance seemed to be very low .
- We suspected that explained variance may partly depend on sample size.
- To test this hypothesis we generated random subsets (10) for 5, 10, 20, 30, 50, 60 % of relevés of the whole data set and analyzed them separately.
- In fact, a clear decrease in explained variance with increasing sample size was observed.

## Eigenvalues in large data sets



# Forward selection of abiotic explanatory variables: soil, climate, spatial

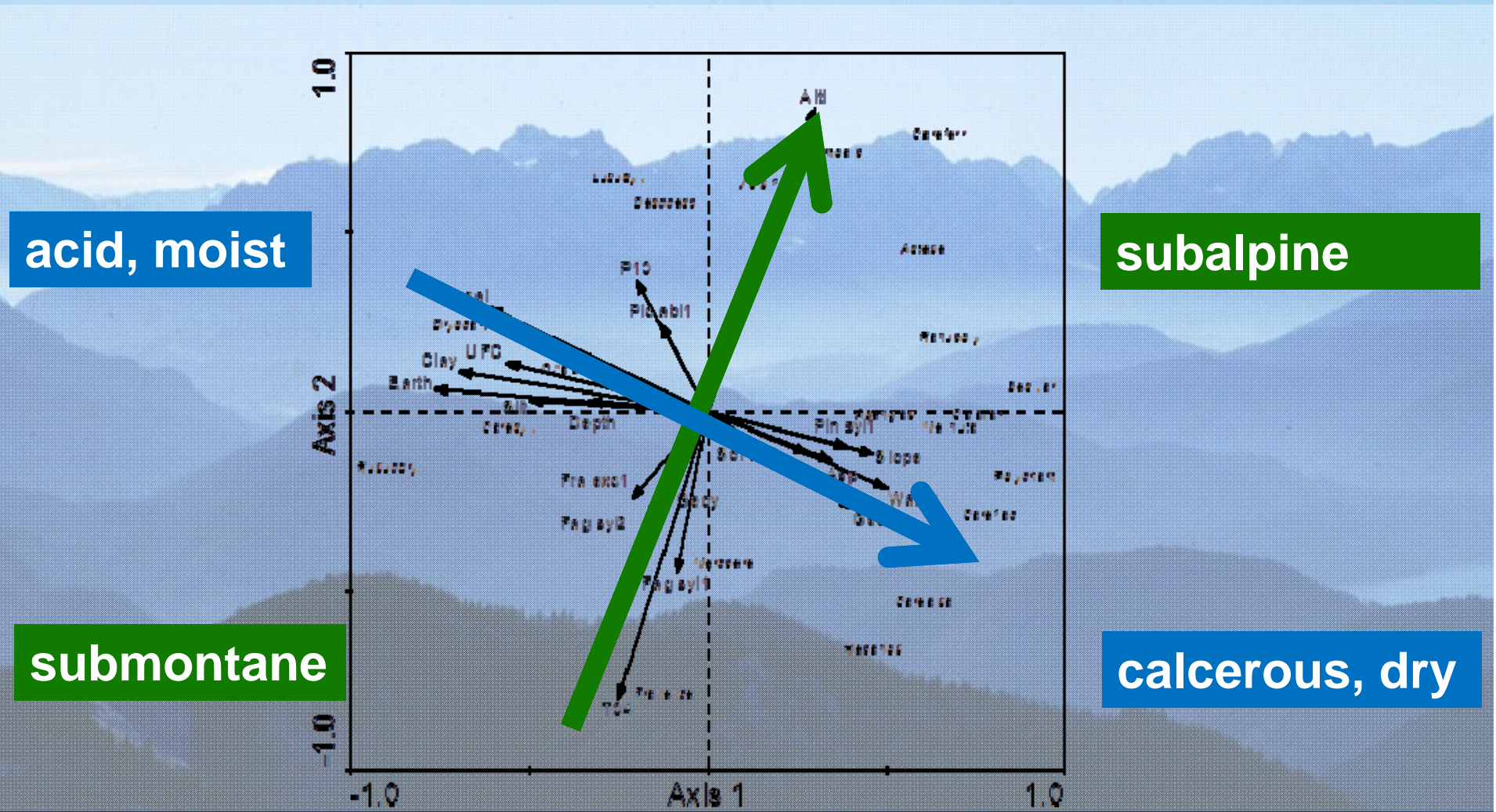
Code	Explanatory Variable	Unit	IQR	Total	Tree	Unders.
<b>Soil</b>						
Org	Thickness of organic layer	cm	[2, 6]	X		X
Ah	Thickness of A <sub>h</sub> -Horizon	cm	[7, 23]	X	X	X
Depth	Depth of soil	cm	[55, 100]	X		X
Earth	Volume of fine earth	l/m <sup>2</sup>	[230, 755]	X	X	X
Clay	Volume of clay	l/m <sup>2</sup>	[33, 201]	X		X
Silt	Volume of silt	l/m <sup>2</sup>	[119, 362]	X	X	X
Sand	Volume of sand (collinear with clay and silk)	l/m <sup>2</sup>	[20, 126]			
Water	Depth of waterlogging	cm	[100, 100]	X	X	X
Decal	Depth of depletion of lime (Decalcification)	cm	[0, 59]	X	X	X
UFC	Useable field capacity	mm	[64, 152]	X		X
<b>Climate</b>						
P06	Sum of precipitation falling in June	mm	[64, 152]		X	
P10	Sum of precipitation falling in October	mm	[107, 134]	X		X
T04	Average temperature in April	°C	[3, 5]	X		X
T05	Average temperature in May	°C	[8, 10]		X	
Alti	Altitude a.s.l.	m	[890, 1220]	X	X	X
Asp	cos (195°-aspect) „southernness“	./.	[-0.84, 0.64]	X		X
Slo	Slope	degree	[15, 35]	X	X	X
<b>Spatial</b>						
Geox	x-coordinate, Gauß-Krüger coordinate system	m		X	X	X
Geoy	y-coordinate, Gauß-Krüger coordinate system	m		X	X	X



# Forward selection of biotic explanatory variables, tree species

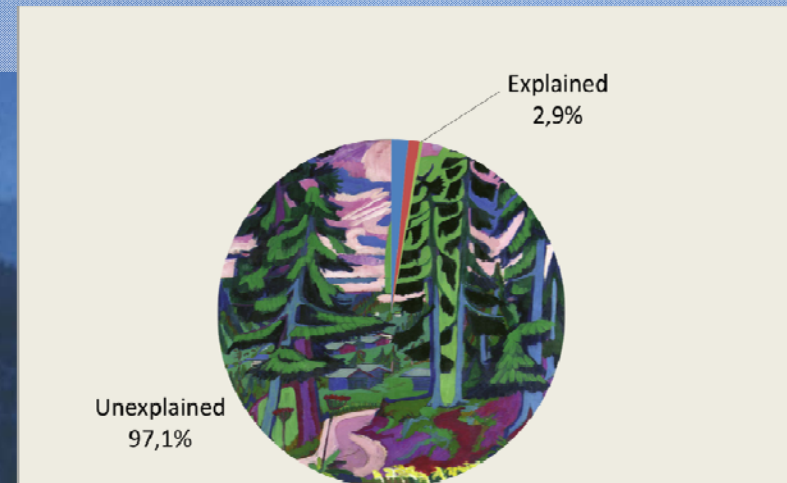
Code	Name	Quantiles [% cover]			Frequency
		Q <sub>25</sub>	Q <sub>50</sub>	Q <sub>75</sub>	
Abialb1	Abies alba, T1	1	5	25	353
Acepse1	Acer pseudoplatanus, T1	1	5	5	314
Aceps2	Acer pseudoplatanus, T2	1	1	5	165
Alnaln1	Alnus alnobetula, T1	26	50	62	3
Alnglu1	Alnus glutinosa, T1	5	15	25	6
Alnglu2	Alnus glutinosa, T2	0	1	1	4
Alninc1	Alnus incana, T1	5	5	25	33
Alninc2	Alnus incana, T2	1	1	5	17
Betpen1	Betula pendula, T1	1	5	5	7
Corave2	Corylus avellana, T2	5	5	5	5
Fagsyl1	Fagus sylvatica, T1	5	25	50	699
Fagsyl2	Fagus sylvatica, T2	1	5	5	484
Fraexc1	Fraxinus excelsior, T1	1	5	25	106
Fraexc2	Fraxinus excelsior, T2	1	5	5	39
Lardec1	Larix decidua, T1	1	1	5	50
Picabi1	Picea abies, T1	5	25	50	1222
Picabi2	Picea abies, T2	1	1	5	550
Pincem1	Pinus cembra, T1	1	1	13	3
Pinmug1	Pinus mugo agg., T1	20	38	50	4
Pinsyl1	Pinus sylvestris, T1	1	5	25	51
Pinsyl2	Pinus sylvestris, T2	1	3	5	12
Poptre1	Populus tremula, T1	1	3	4	2
Prupad2	Prunus padus, T2	5	5	5	2
Querob1	Quercus robur, T1	5	5	5	3
Salapp1	Salix appendiculata agg., T1	2	3	4	2
Salarb1	Salix arbuscula agg., T1	7	13	9	2
Sorari1	Sorbus aria agg., T1	1	1	5	17
Sorari2	Sorbus aria agg., T2	1	1	1	47
Tilpla1	Tilia platyphyllos, T1	5	25	25	5
Ulmgl2	Ulmus glabra, T2	1	1	5	32

# Gradients underlining species composition of the understorey



# Variance partitioning

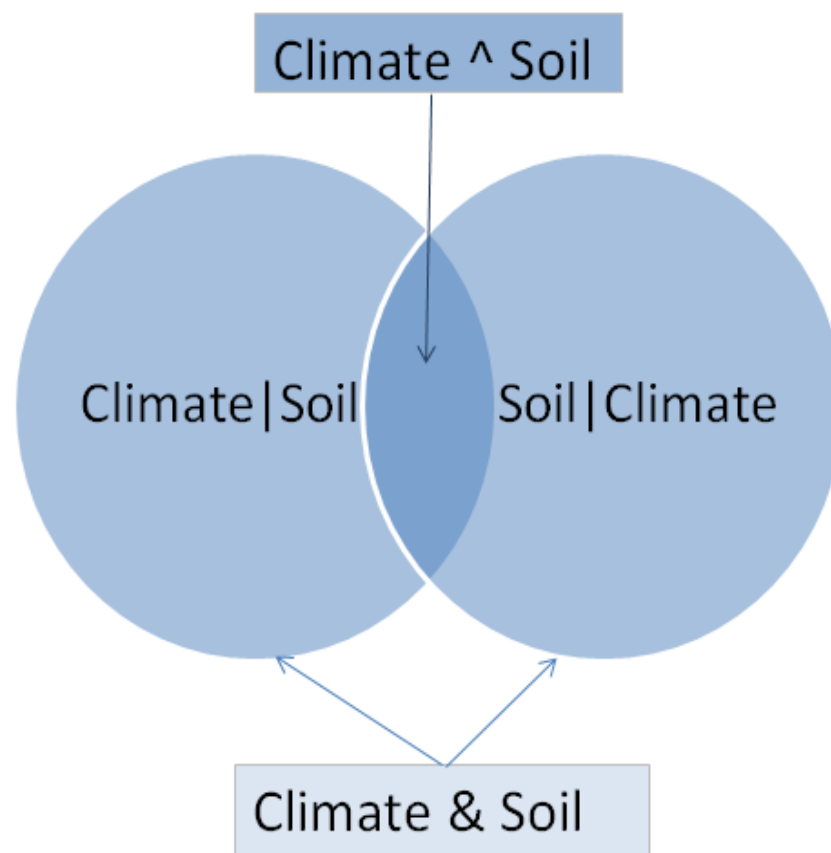
- Canonical analyses allow partitioning the observed variability of species compositions into a portion that can be explained by the observed environmental variables and into unexplained remains.



# Variance partitioning

- Using partial ordination, the explained variability can be partitioned according to logical groups of environmental variables.
- We grouped the predictor variables into abiotic: “soil”, “climate”, “spatial” and into biotic: “tree layer” .

# Partitioning model



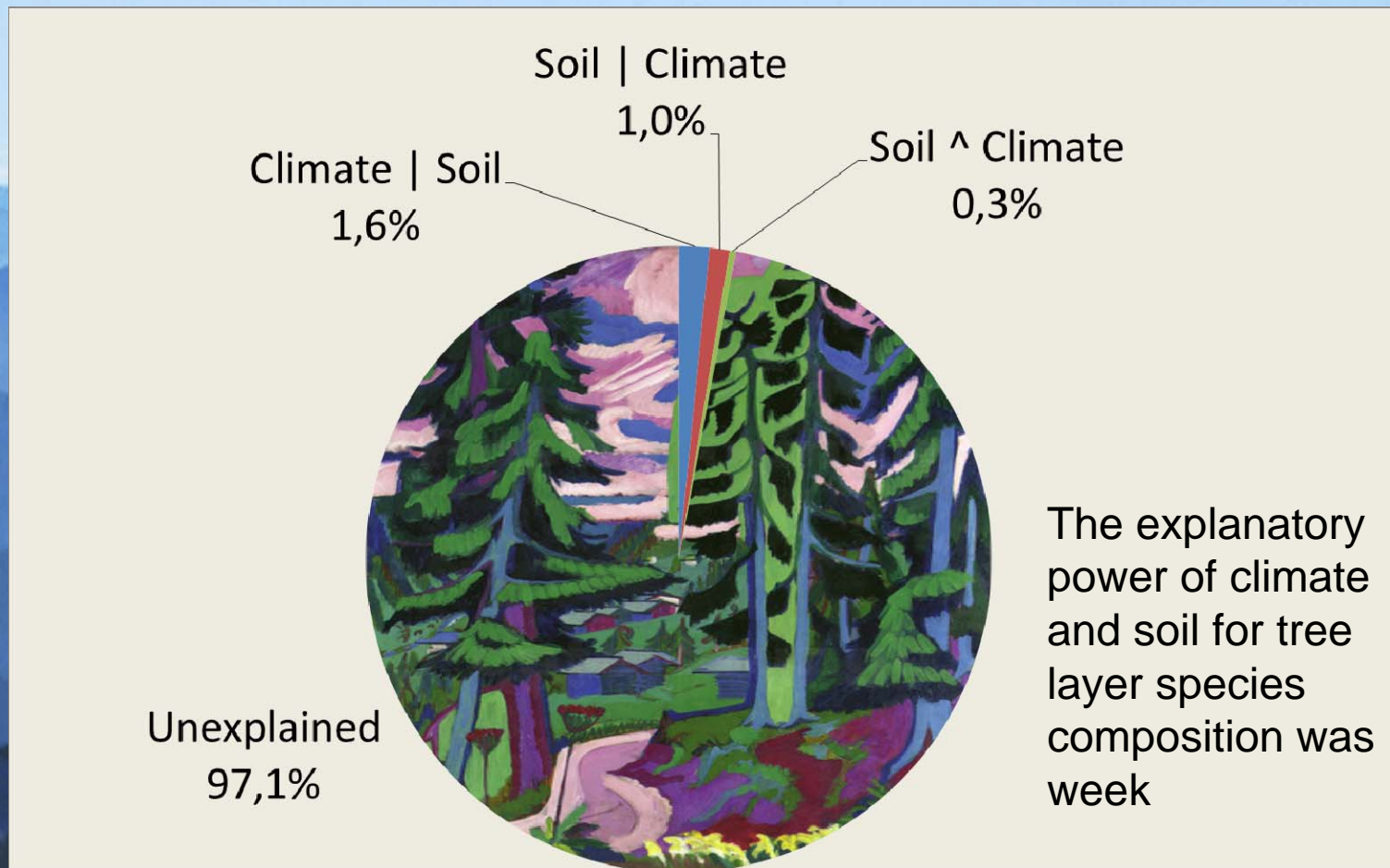
e.g.

- **Climate & Soil: total variance explained by both**
- **Climate | Soil: variance explained by climate only**
- **Soil | Climate: variance explained by soil only**
- **Climate ^ Soil: variance explained by soil AND climate**

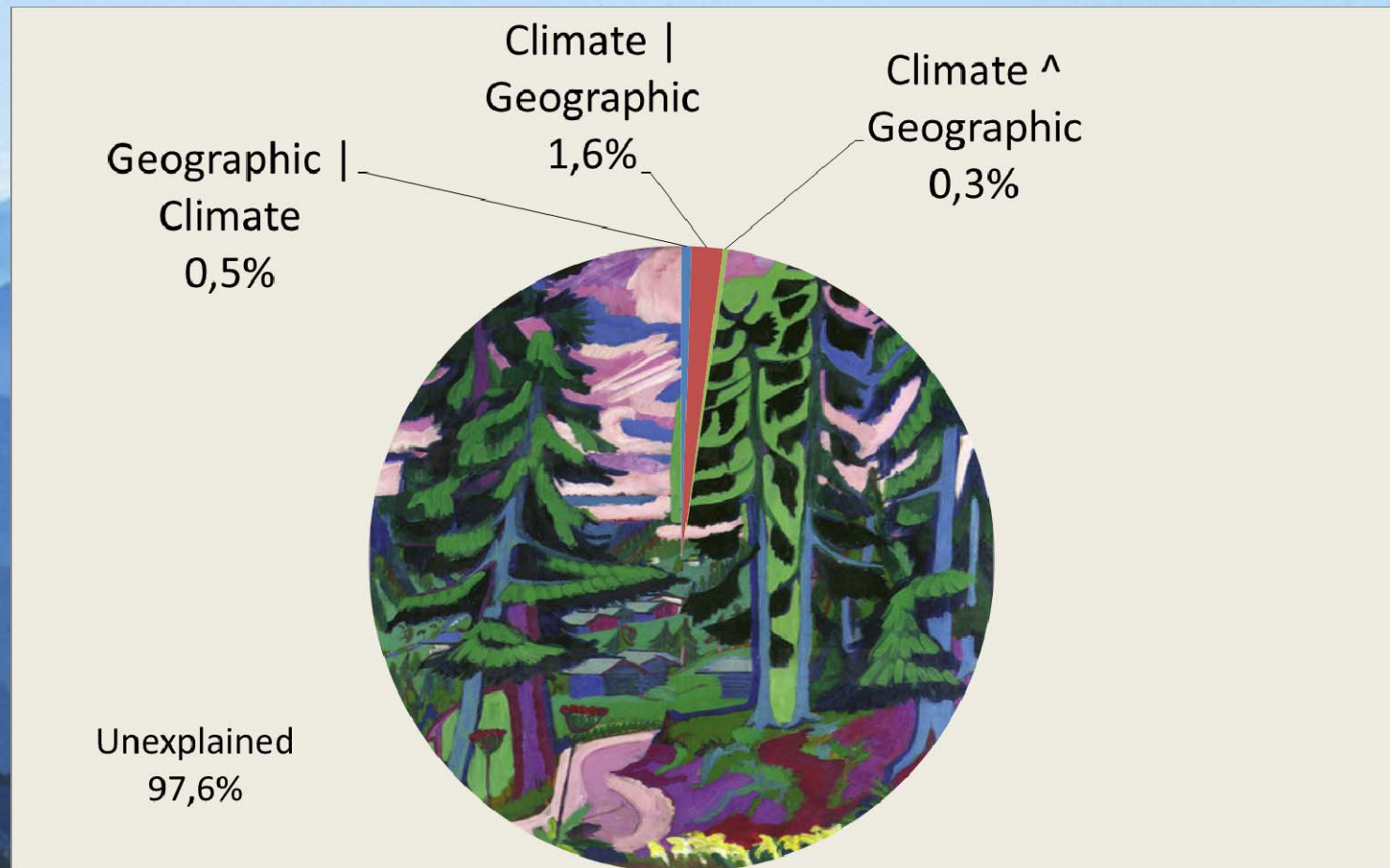
note

**Climate & Soil =  
Climate | Soil + Climate ^ Soil  
+ Soil | Climate**

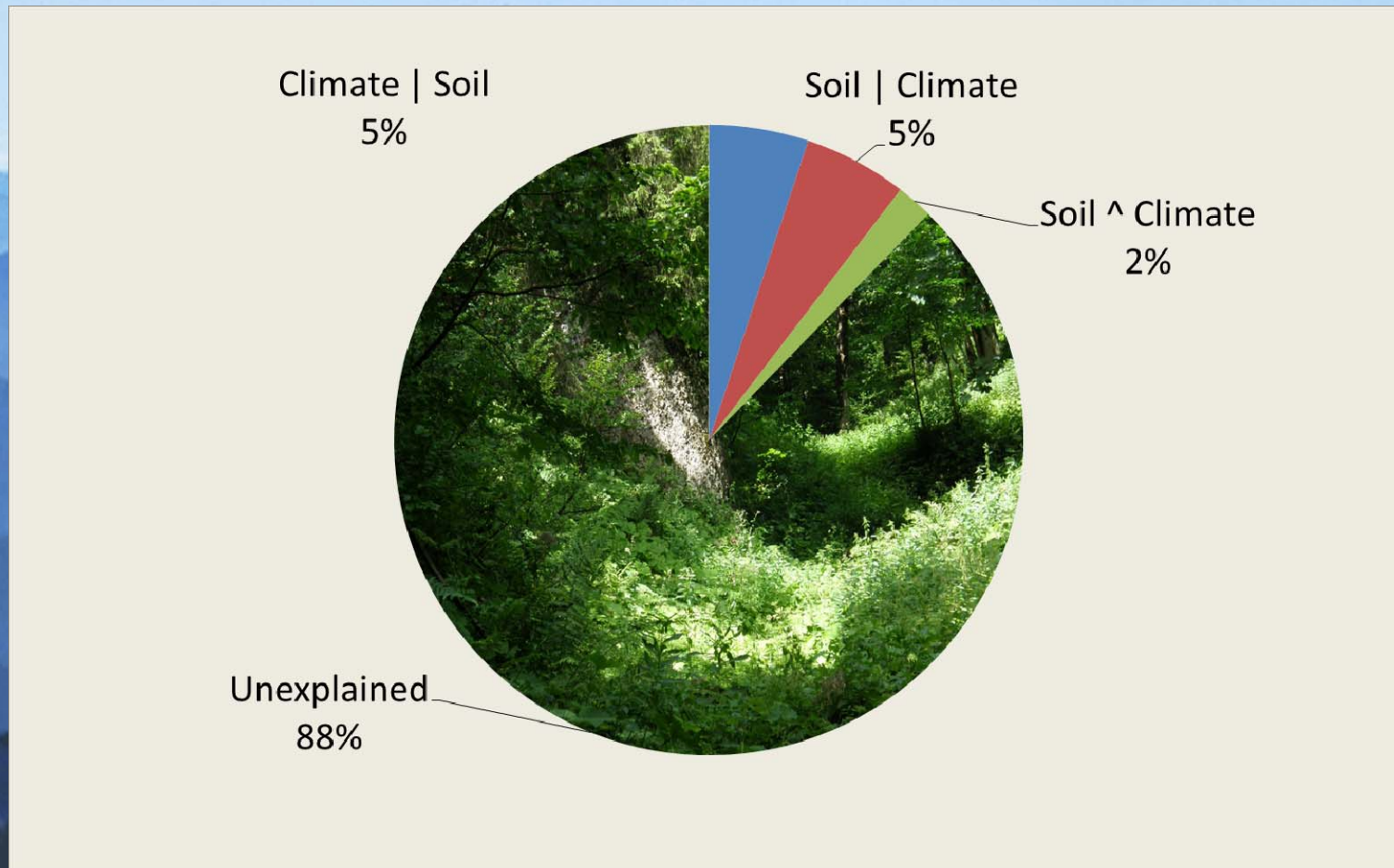
# Variance partitioning tree layer: climate and soil



# Variance partitioning tree layer: geographic and climate

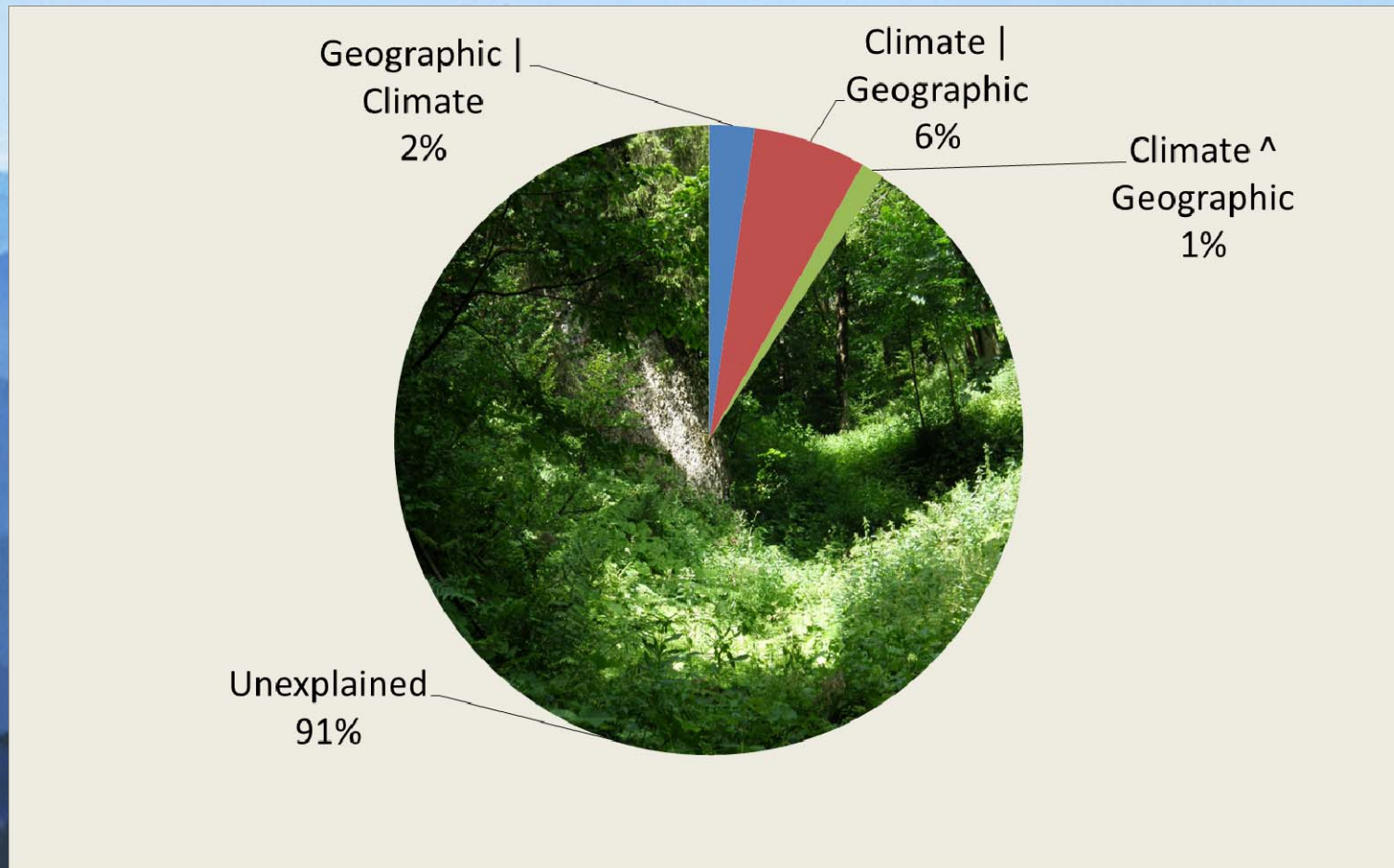


# Variance partitioning understory: climate and soil

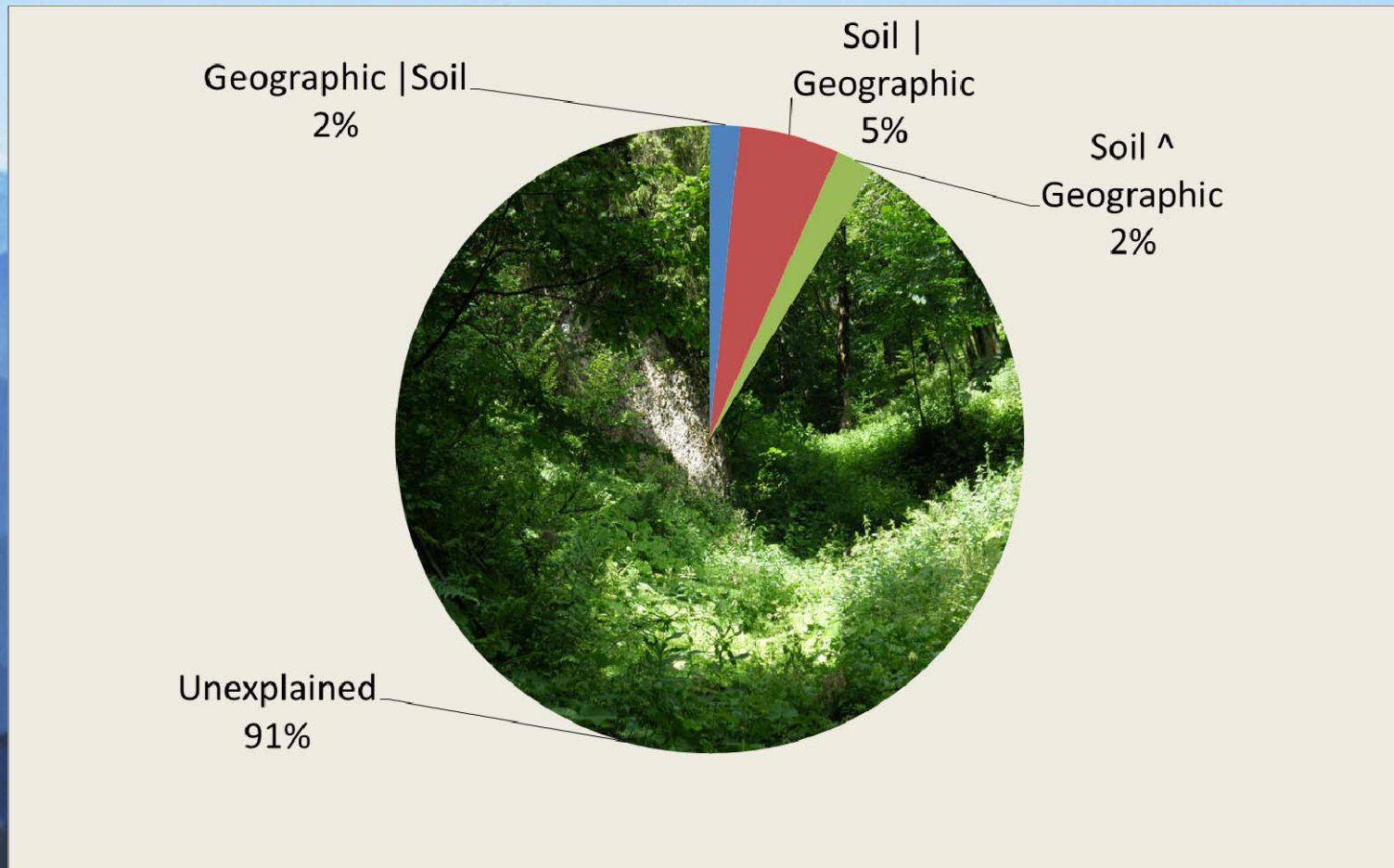




# Variance partitioning understory: geographic and climate



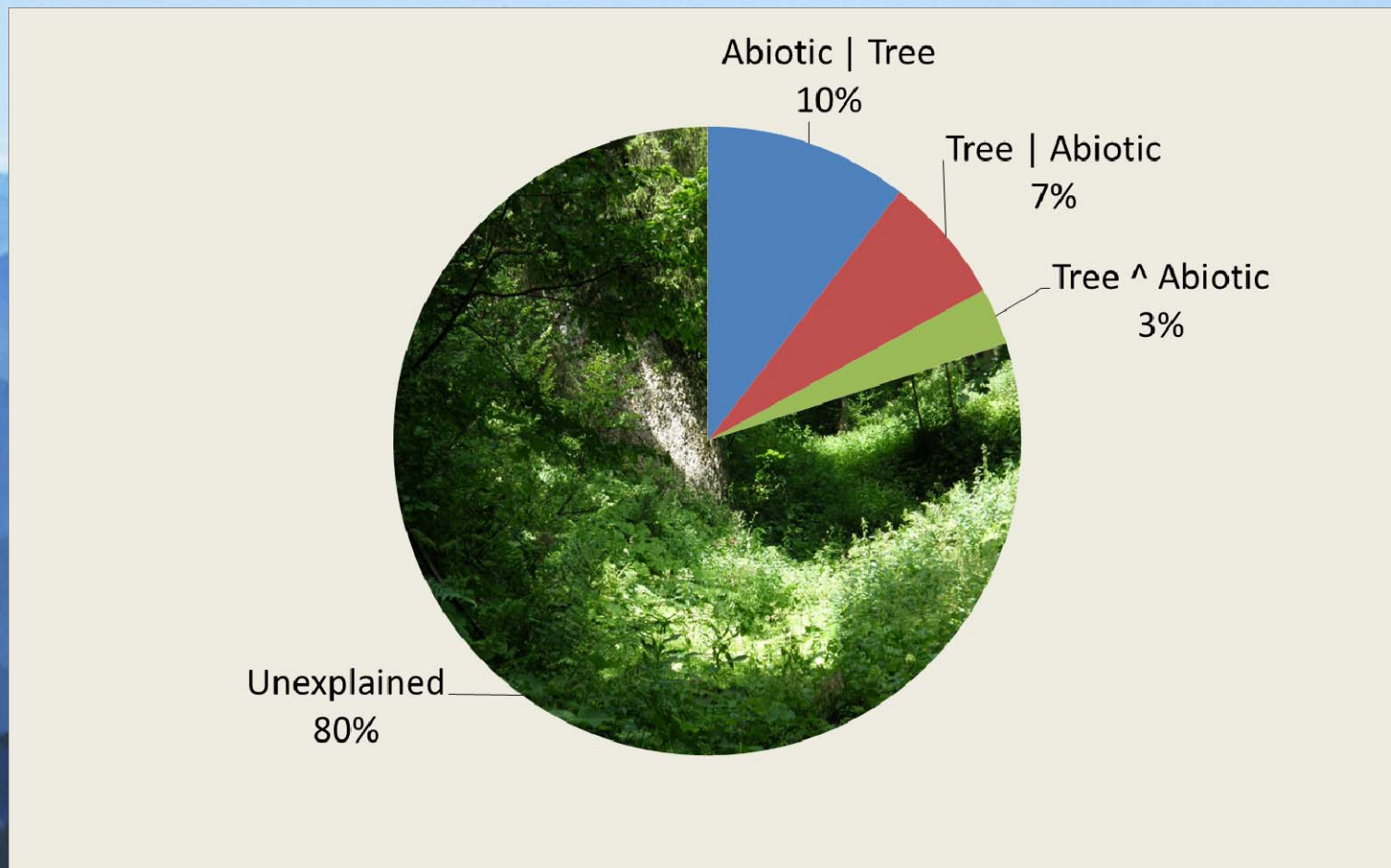
# Variance partitioning understory: geographic and soil



# Variance partitioning understory: abiotic and trees

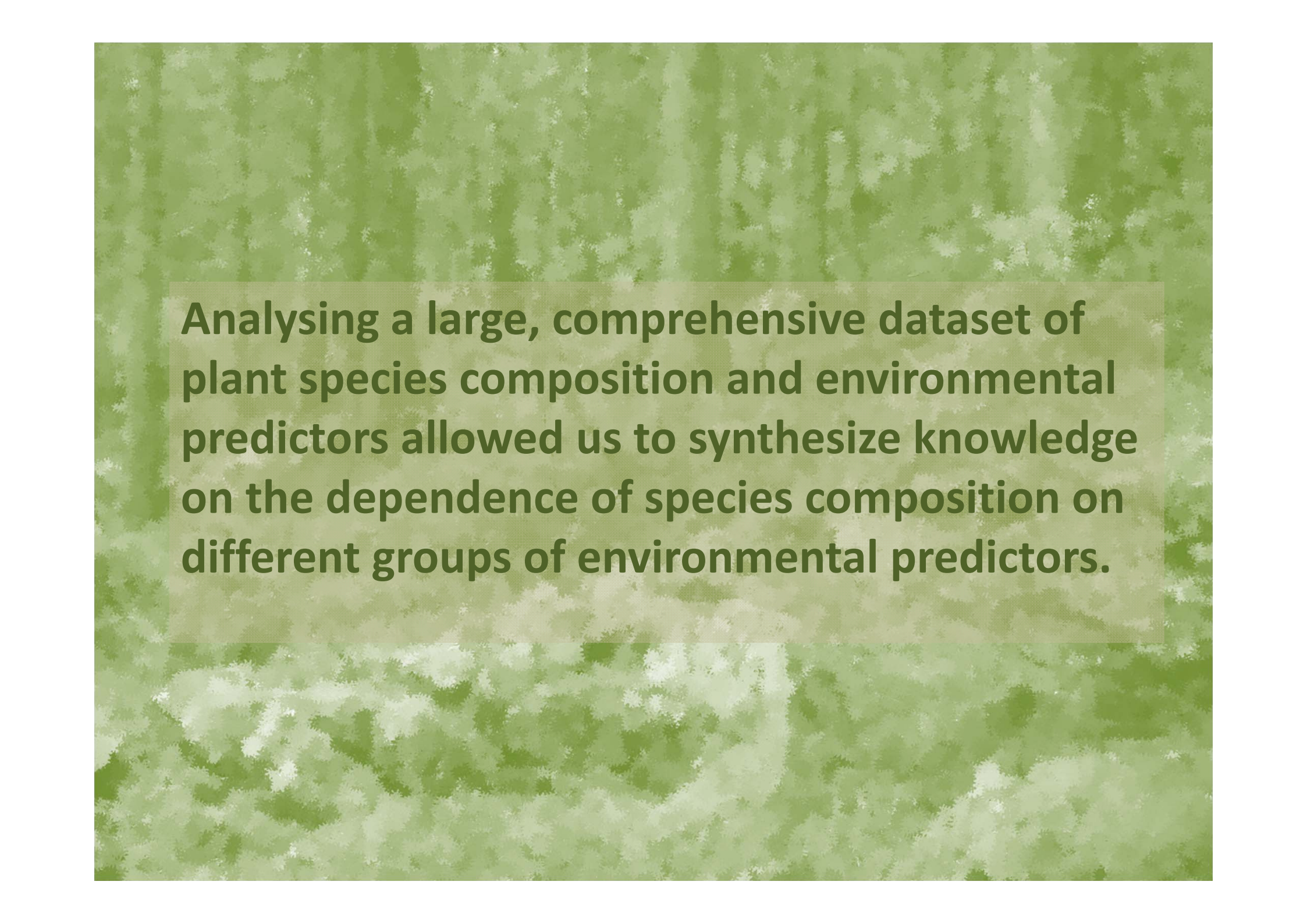
- We hypothesized that the tree layer influences the environment of understory vegetation.
- Therefore we tested tree species as one set of predictor variables against all abiotic variables.
- The largest portion of explained variability (10 %) was associated with the abiotic environmental variables.
- But tree layer had an additional explanatory power of 7 %.

# Variance partitioning understory: abiotic and trees



# Conclusion

- **Understory and tree layer react different to environmental variables.**
- **Understory is better explained by choosen environmental variables than tree layer.**
- **Understory: Climate and soil explain nearly equal portion of variance.**
- **Understory: Tree species (Biotic) explain huge part (7%) of the variability of the understorey species composition.**



**Analysing a large, comprehensive dataset of plant species composition and environmental predictors allowed us to synthesize knowledge on the dependence of species composition on different groups of environmental predictors.**

# We would like to express our great appreciation to

## WINALP-Project-team and

- Bayerische Vermessungsverwaltung (LVG)
- Landesamt für Umwelt (LfU)
- Deutscher Wetterdienst (DWD)
- Landesanstalt für Wald und Forstwirtschaft (LWF)





WINALP



**Thank you for your attention!**