

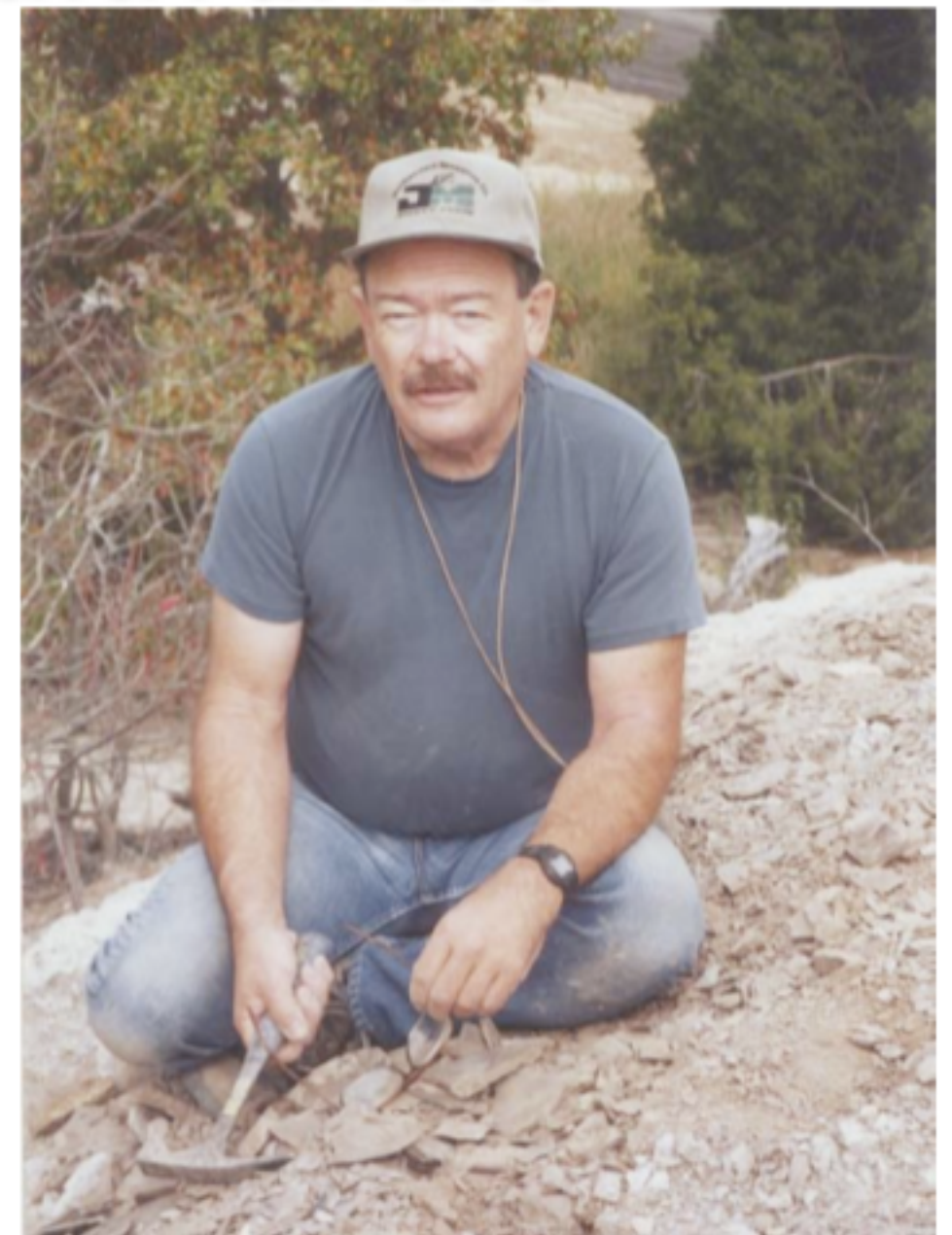


# CLAMP

## *Climate Leaf Analysis Multivariate Program*

*Bob Spicer (r.a.spicer@open.ac.uk)*

*A palaeoclimate proxy based on leaf physiognomy initially developed by the late Jack A Wolfe.*



Jack a Wolfe

# Plant Palaeoclimate Proxies

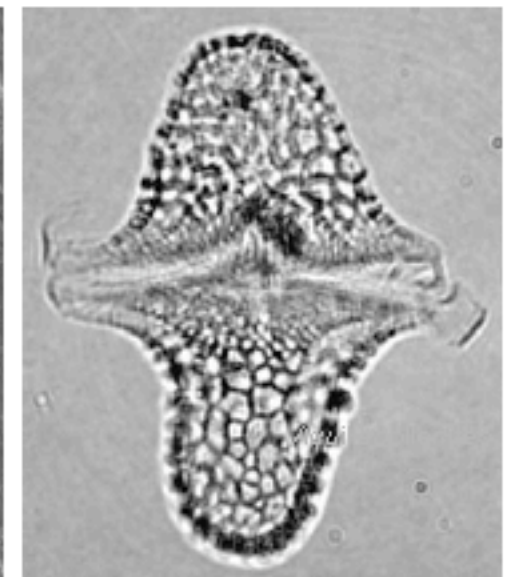


Palaeoclimate proxies may be divided into two types:

- 1) those based on the environmental tolerances of assumed living relatives (nearest living relative approaches) and
- 2) those that are based on aspects of plant architecture constrained by environmental conditions (physiognomic approaches).

Each have specific advantages and disadvantages.

**Nearest living relative techniques** (NLR) can be applied to all plant organs assignable to modern taxa but are most useful for those plant organs lacking known morphological adaptations to the physical environment (e.g. seeds and pollen). However they are restricted to timescales where evolutionary change at the species level is unlikely. In most cases <1-5Ma, although techniques that involve protocols for examining populations (e.g. Co-existence Analysis and Overlapping Distribution Analysis) can be extended further back in time.





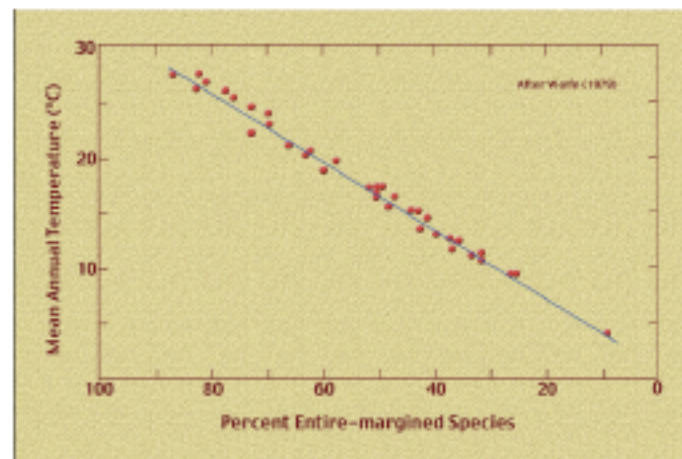
# Plant Palaeoclimate Proxies

**Physiognomic** - Climatic signals encoded in plant architecture as a developmental and growth response to the environment, honed by selection to maximise functional efficiency. Examples - wood anatomy, leaf architecture and stomatal analysis.

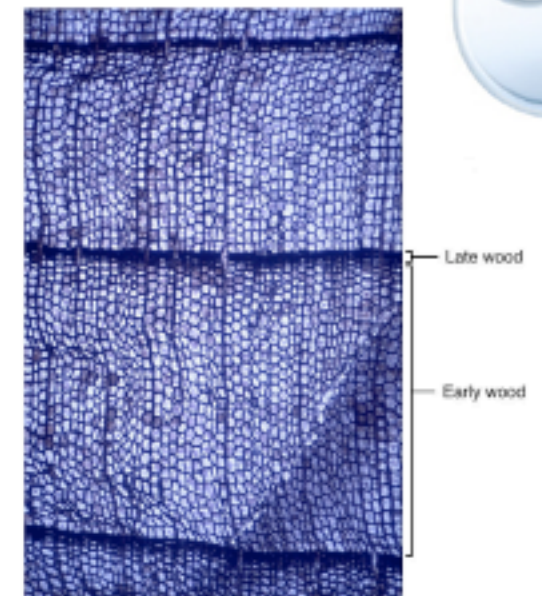
These techniques have the advantage of being useful over long timescales that encompass previous greenhouse climates, but are restricted to comparatively rare leaf and wood assemblages.



Stomatal analysis can yield estimates of  $\text{ppCO}_2$  and, in ideal circumstances, estimates of elevation.

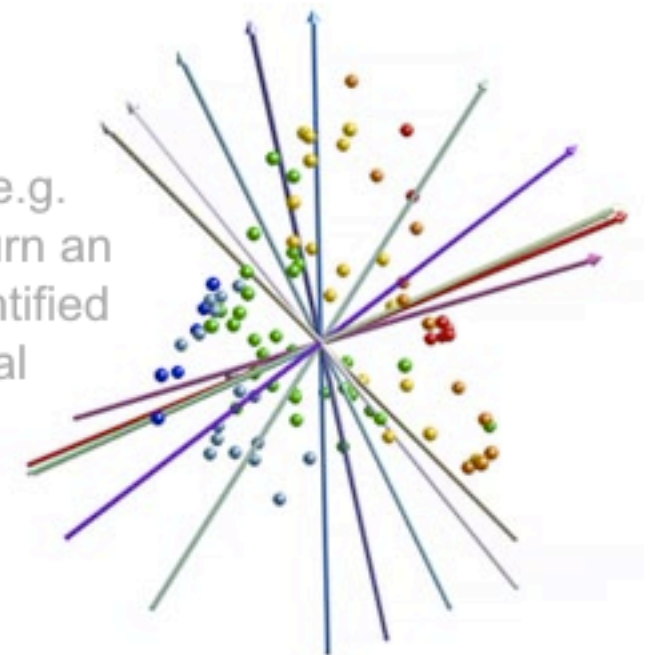


Univariate Techniques - Leaf Margin Analysis - return a single variable e.g. mean annual temperature.



Wood records almost daily variations in growth conditions but is difficult to quantify unambiguously in terms of climate variables.

Multivariate techniques (e.g. CLAMP) return an array of quantified environmental variables.







In desert regions where water is in short supply leaves are either small or have been dispensed with altogether in favour of photosynthetic stems. Additional adaptations include stem enlargement for water storage.



These adaptations are universal and are governed by the physics of evaporation. A small surface area to volume ratio is advantageous in limiting evaporation.

Because this represents a general “engineering solution” to the problem of water loss similar morphologies have evolved in similar environments independent of taxonomic affiliations.



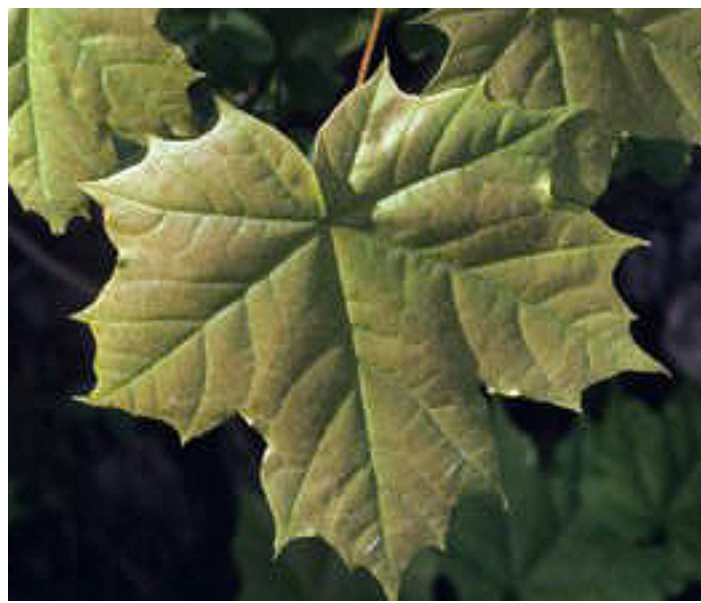




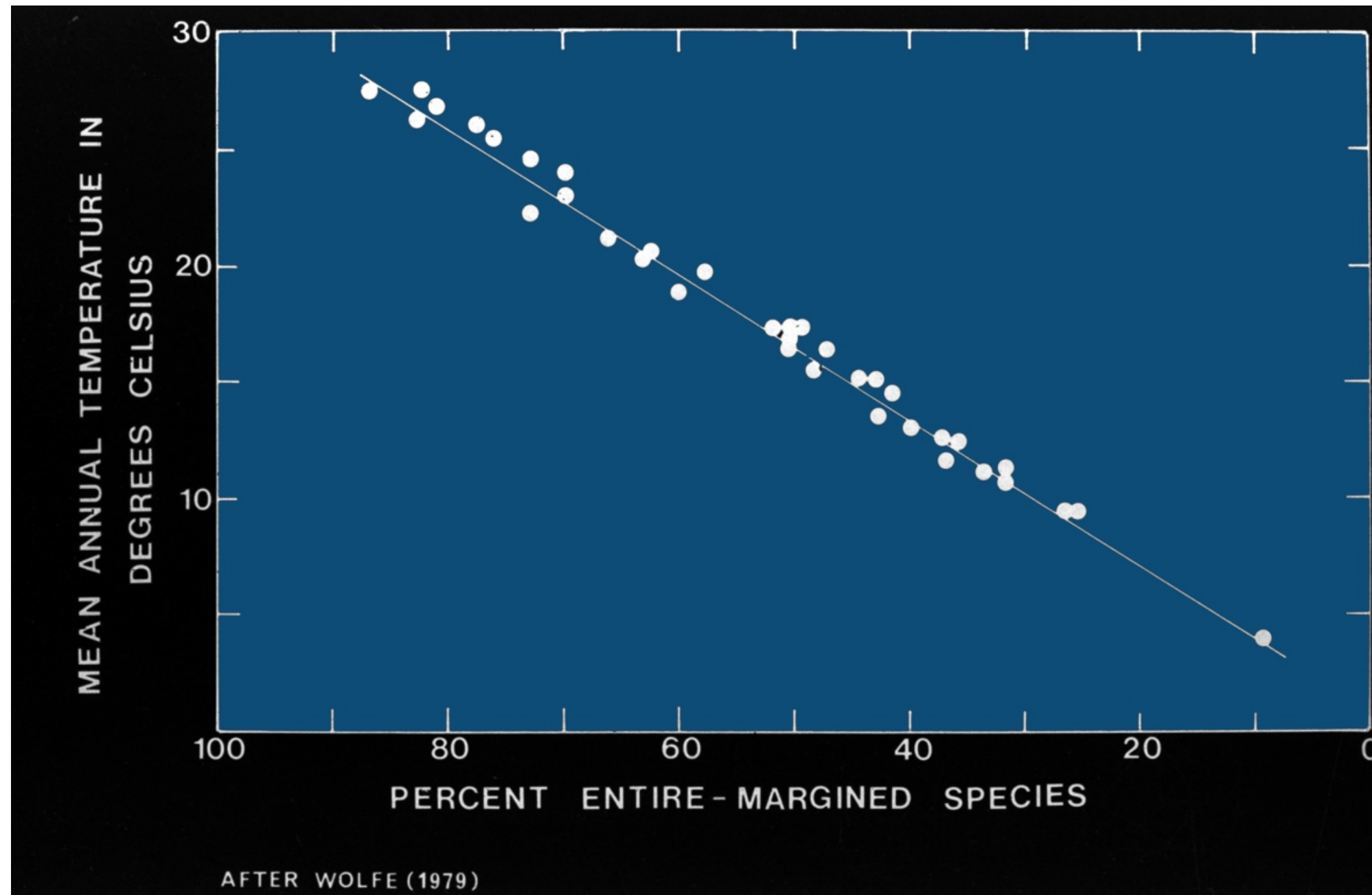
In environments where water supply is not limiting a much larger leaf surface area can be supported without desiccation. Leaf area indices (total leaf area/unit area of ground surface)  $> 12$  can be observed in rain forests. Individual leaf size range can be large.

However leaves also reflect the local microclimate: leaves at the top of the tree crown are exposed to high insolation and wind speeds so are smaller and thicker than leaves in the darker, more humid, understory.





Other aspects of leaf architecture vary with environment. As long ago as 1915 Bailey and Sinnott noted in *Science* that for woody dicots in North America the proportion of taxa bearing toothed versus entire margined leaves varies with mean annual temperature.



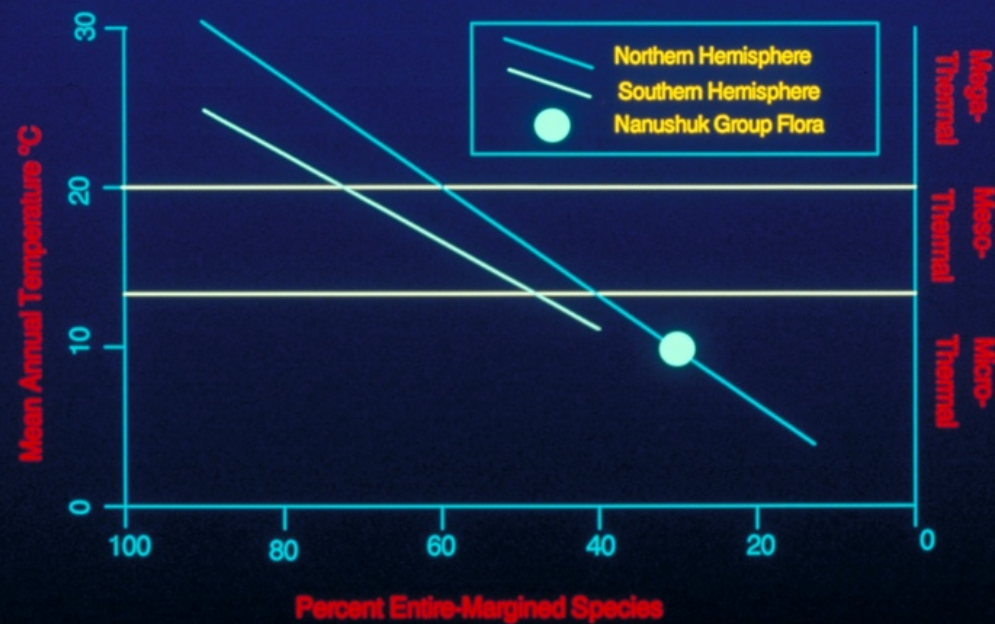
In 1979 Wolfe retested this relationship using leaves from S.E. Asia.

The relationship only works where water is not limiting, i.e. in humid to mesic forests.





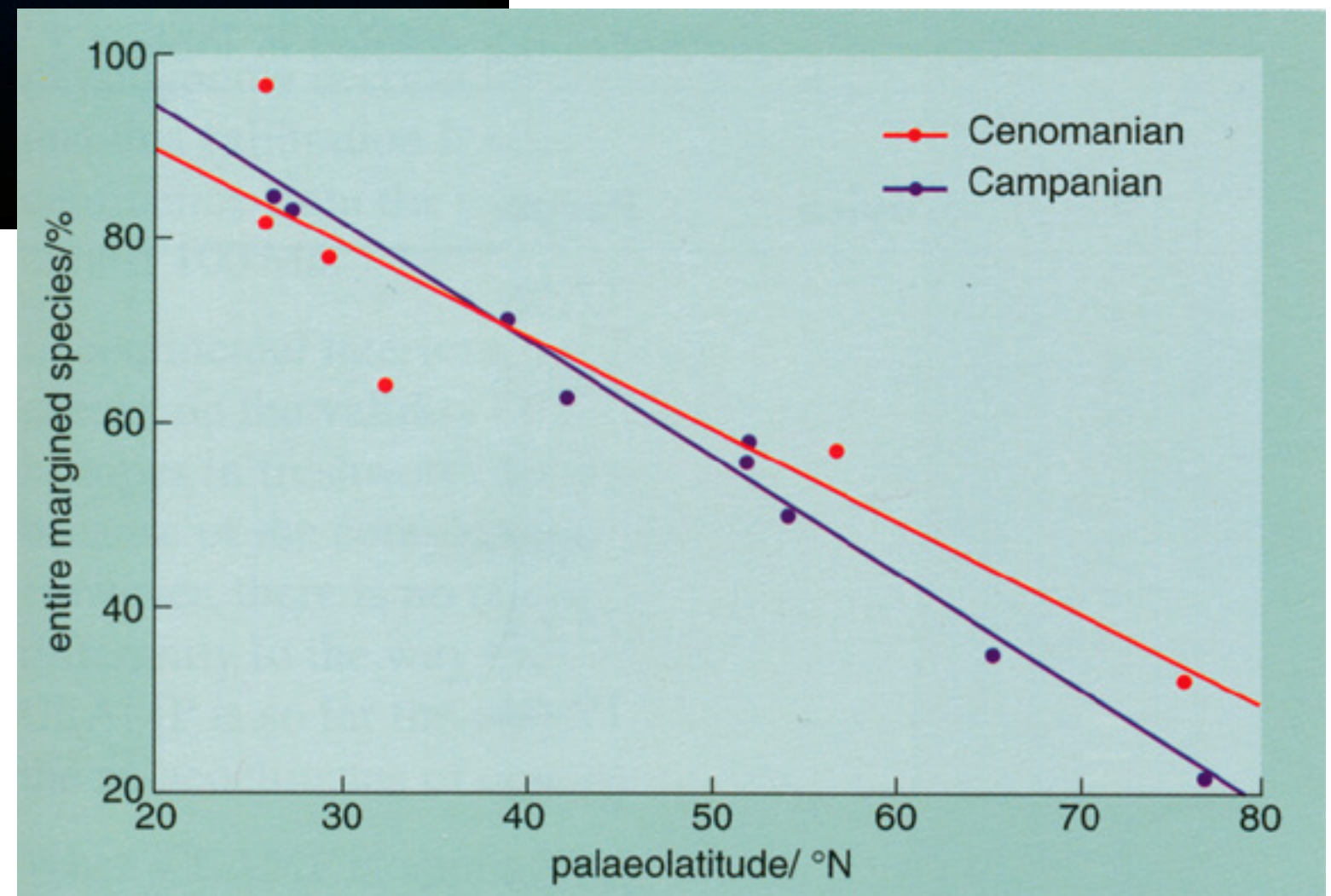
## Leaf Margin Type and Temperature



Wolfe also noted that the relationship, as evidenced by the slope of the regression line, differs between Northern and Southern Hemispheres.

When tested using fossil floras Wolfe also noted that the percentage of entire margined taxa when plotted against palaeolatitude tracks changes in equator-to-pole temperature gradients (and global mean surface temperature).

There appears to be no change in slope associated with the polar light regime ( $>66^\circ$ ).

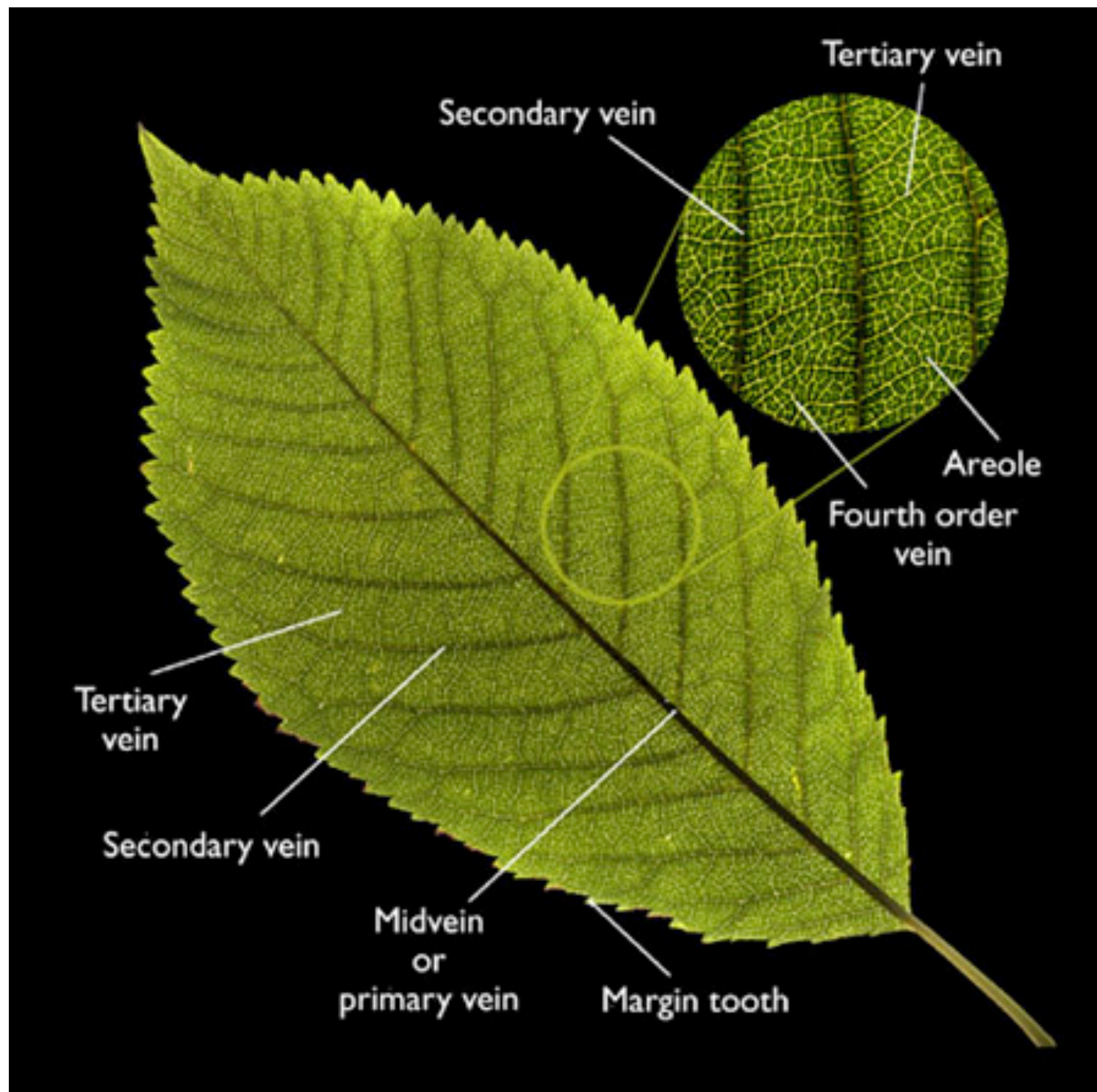


Wolfe also noted that the relationship, as evidenced by the slope of the regression line, differs between Northern and Southern Hemispheres. This relationship has been revisited and refined by others (e.g. Wilf, P. 1997. When are leaves good thermometers? A new case for Leaf Margin Analysis, *Paleobiology* 23 : 373-390). The underlying cause is unclear, but it is likely to involve several factors including maintaining leaf temperature and fluid flow within the plant. As with all leaf architectural features there is no simple relationship between any given feature and any given climate parameter.

When tested using fossil floras Wolfe also noted that the percentage of entire margined taxa when plotted against palaeolatitude tracks changes in equator-to-pole temperature gradients (and global mean surface temperature) even in the late Cretaceous when thermal equator to pole gradients are thought to have been shallower than today.

There appears to be no change in slope associated with the polar light regime ( $>66^\circ$ ).



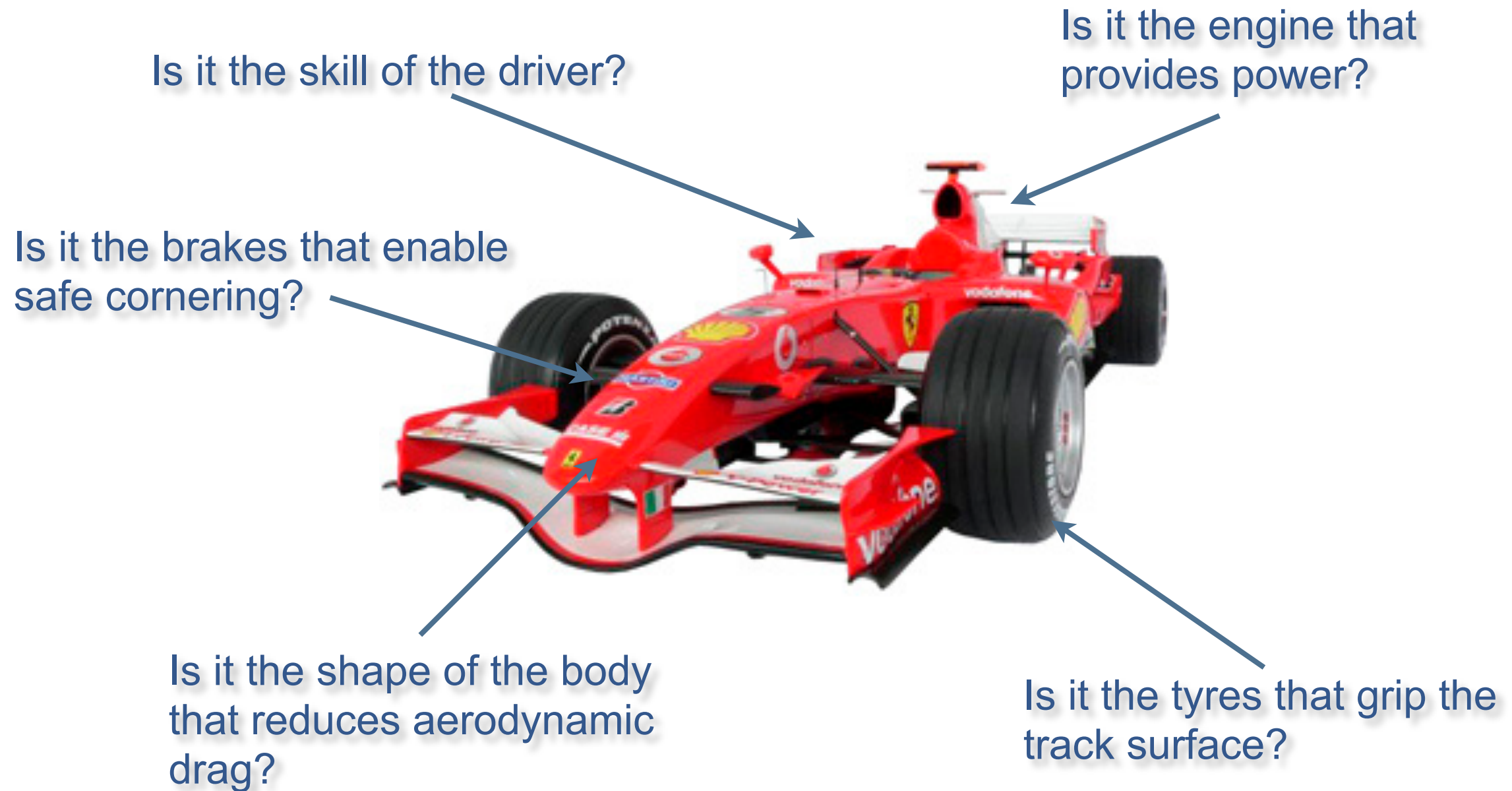


Wolfe surmised that if leaf size was primarily related to water loss and margin characteristics were primarily related to mean annual temperature, then it was likely that many other leaf architectural characteristics might carry environmental signals.

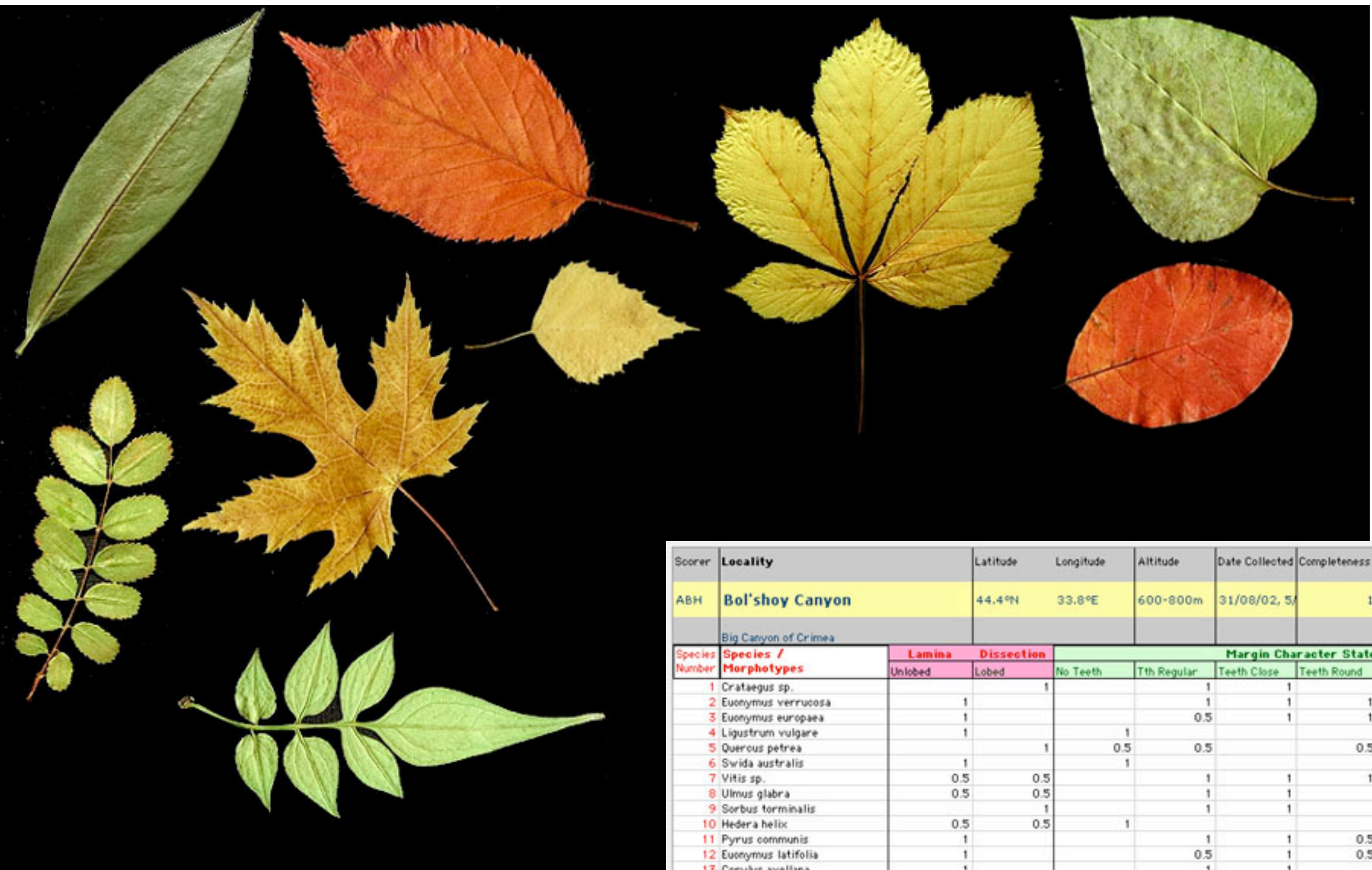
However it is unlikely that each architectural (physiognomic) feature relates to a separate environmental parameter (variable).



# What feature of an F1 racing car make it capable of outcompeting other F1 cars in a race?







Through experiment Wolfe identified 31 physiognomic character states that could be scored for multivariate analysis encompassing lobing, margin characteristics, size, apex and base form and shape.

Scorer	Locality	Latitude	Longitude	Altitude	Date Collected	Completeness	Notes		
ABH	Bol'shoy Canyon	44.4°N	33.8°E	600-800m	31/08/02, 5/	1			
	Big Canyon of Crimea								
Species Number	Species / Morphotypes	Lamina	Dissection	Margin Character States					
		Unlobed	Lobed	No Teeth	Tth Regular	Teeth Close	Teeth Round	Teeth Acute	Tth Co
1	Crataegus sp.		1		1	1		1	
2	Euonymus verrucosa	1			1	1	1		
3	Euonymus europaea	1			0.5	1	1		
4	Ligustrum vulgare	1		1					
5	Quercus petraea		1	0.5	0.5		0.5		
6	Swida australis	1		1					
7	Vitis sp.	0.5	0.5		1	1	1		
8	Ulmus glabra	0.5	0.5		1	1		1	
9	Sorbus torminalis		1		1	1		1	
10	Hedera helix	0.5	0.5	1					
11	Pyrus communis	1			1	1	0.5	0.5	
12	Euonymus latifolia	1			0.5	1	0.5	0.5	
13	Corylus avellana	1			1	1		1	
14	Fagus silvestris	1			1		1		
15	Carpinus orientalis	1			1	1		1	
16	Sambucus nigra	1			1	1		1	
17	Berberis vulgaris	1			0.5			1	
18	Cornus mas	1		1					
19	Cotinus coggygria	1		1					
20	Tilia cordata	1			1	1		1	
21	Euphorbia amygdaloides	1		1					
22	Rosa canina	1			1	0.5		1	
23	Salix sp.	1			0.5	0.5	0.5	0.5	
24	Carpinus betulus	1			1	1		1	
25	Clematis vitalba	0.5	0.5		0.5	0.5	0.5	0.5	
26	Acer campestre		1	0.5	0.25	0.25	0.5		

Through experiment Wolfe identified 31 physiognomic character states that could be scored for multivariate analysis encompassing lobing, margin characteristics, size, apex and base form and shape.





# A typical CLAMP scoresheet

Bereyozovo correct.xls [Read-Only]													
	A	B	C	D	E	F	G	H	I	J	K	L	M
1	Scorer	Locality			Latitude	Longitude	Altitude	Date Collected	Completeness	Notes			
2													
3	ABH	Beryozovo			63° 54.469N	64° 56.707E	63'	1/8/02	1				
4													
5	Species	Species /	Lamina	Dissection	Margin Character States								
6	Number	Morphotypes	Unlobed	Lobed	No Teeth	Tth Regular	Teeth Close	Teeth Round	Teeth Acute	Tth Compound	Nanophyll	Leptophyll I	Leptophyll II
7	1	"Bilberry" sp.1	1			0.5	1	0.5	0.5		0.33	0.33	0.33
8	2	Sorbus	1			1	0.5		1	0.5			0.5
9	3	Populus tremula	1				0.5	0.5	0.5	0.5			0.25
10	4	Rosa sp.	1			1	1	0.5	0.5	0.5	0.2	0.2	0.2
11	5	Vaccinium vitis-idaea	1		1						0.25	0.25	0.25
12	6	Alnus sp.	1			1	1		1	1			
13	7	Salix sp. 4	1		1							0.2	0.2
14	8	Salix sp. 2	1		1							0.33	0.33
15	9	Vaccinium uliginosum	1		1						0.25	0.25	0.25
16	10	Unknown sp. 1	1		0.5		0.5		1		0.2	0.2	0.2
17	11	Unknown sp. 2	1		1							0.33	0.33
18	12	Empetrum sibiricum	1		1						1		
19	13	Betula sp. 1*	1			1	1		1	1			
20	14	Betula verrucosa *	1				1	1	1	1			
21	15	Betula nana	1			1	1	1		0.5	0.25	0.25	0.25
22	16	Salix sp. 1	1			1		1			0.25	0.25	0.25
23	17	Salix sp. 3	1		0.5			1				0.25	0.25
24	18	Salix sp. 5	1		0.5			1				0.25	0.25
25	19	Ledum palustre	1		1						0.33	0.33	0.33
26	20	Cranberry	1		1							1	
27													
28													
110		No. Character States Present		20	20	20	20	20	20	20	20	20	20
111													
112		Percentage Score		0	48	38	38	28	33	25	15	22	25
113													
114													

Friday, 24 April 2009

These character states are coded according to strict protocols and entered into a scoresheet like this one that also identified the location of the site (latitude, longitude and altitude), the person who scored it, the date it was collected and how complete it is. In the case of a living site this will usually be 1, but a fossil assemblage may be missing some features for some morphotypes. In this case the completeness is less than one. Any site where the completeness score falls below 0.66 should be discarded because the climate estimates will have a large error associated with them. Also a minimum of 20 taxa (morphotypes) of woody dicot leaves are required to get a reliable CLAMP climate estimate. Almost all the calibration files meet this criteria. The overall score for any given site is summarized by the Percentage Score shown highlighted in pink on the scoresheet.





# A typical CLAMP scoresheet

xls [Read-Only]				xls [Read-Only]									
W	X	Y	Z	AA	AB	AC	AD	AE	AF	AG	AH	AI	AJ
Base Character States				Length to Width Character States					Shape Character States				
x Atten.	Base Cordate	Base Round	Base Acute	L:W<1:1	L:W 1-2:1	L:W 2-3:1	L:W 3-4:1	L:W>4:1	Obovate	Elliptic	Ovate		
		0.5	0.5		0.5	0.5				1			
		0.5	0.5			0.33	0.33	0.33		1			
	0.33	0.33	0.33	0.5	0.5				0.5	0.5			
		0.5	0.5		0.5	0.5				1			
		0.5	0.5		0.5	0.5			0.5	0.5			
		0.5	0.5		1					1			
			1					1		1			
		0.5	0.5			0.33	0.33	0.33		1			
			1		0.33	0.33	0.33		0.5	0.5			
			1		0.33	0.33	0.33			1			
			1			0.5	0.5		0.5	0.5			
		0.5	0.5				0.5	0.5		1			
		0.5	0.5	0.5	0.5					0.5	0.5		
		0.5	0.5	0.5	0.5					0.5	0.5	0.5	
		1		0.5	0.5				0.5	0.5			
			1		0.25	0.25	0.25	0.25		1			
			1			0.33	0.33	0.33		1			
			1			0.5	0.5			1			
		0.5	0.5					1		1			
		1				0.5	0.5			1			
20	20	20	20	20	20	20	20	20	20	20	20		
0	2	37	62	10	27	25	20	19	13	83	5	Copy this line into the foss section of the reference dataset spreadsheet.	

Friday, 24 April 2009

These character states are coded according to strict protocols and entered into a scoresheet like this one that also identified the location of the site (latitude, longitude and altitude), the person who scored it, the date it was collected and how complete it is. In the case of a living sie this will usually be 1, but a fossil assemblage may be missing some features for some morphotypes. In this case the completeness is less than one. Any site where the completeness score falls below 0.66 should be discarded because the climate estimates will have a large error associated with them. Also a minimum of 20 taxa (morphotypes) of woody dicot leaves are required to get an reliable CLAMP climate estimate. Almost all the calibration files meet this criteria. The overall score for any given site is summarized by the Percentage Score shown highlighted in pink on the scoresheet.



A CLAMP calibration matrix consists of a number of modern vegetation sites each with a numerical description of the leaves found on at least 20 woody dicot species in each site.



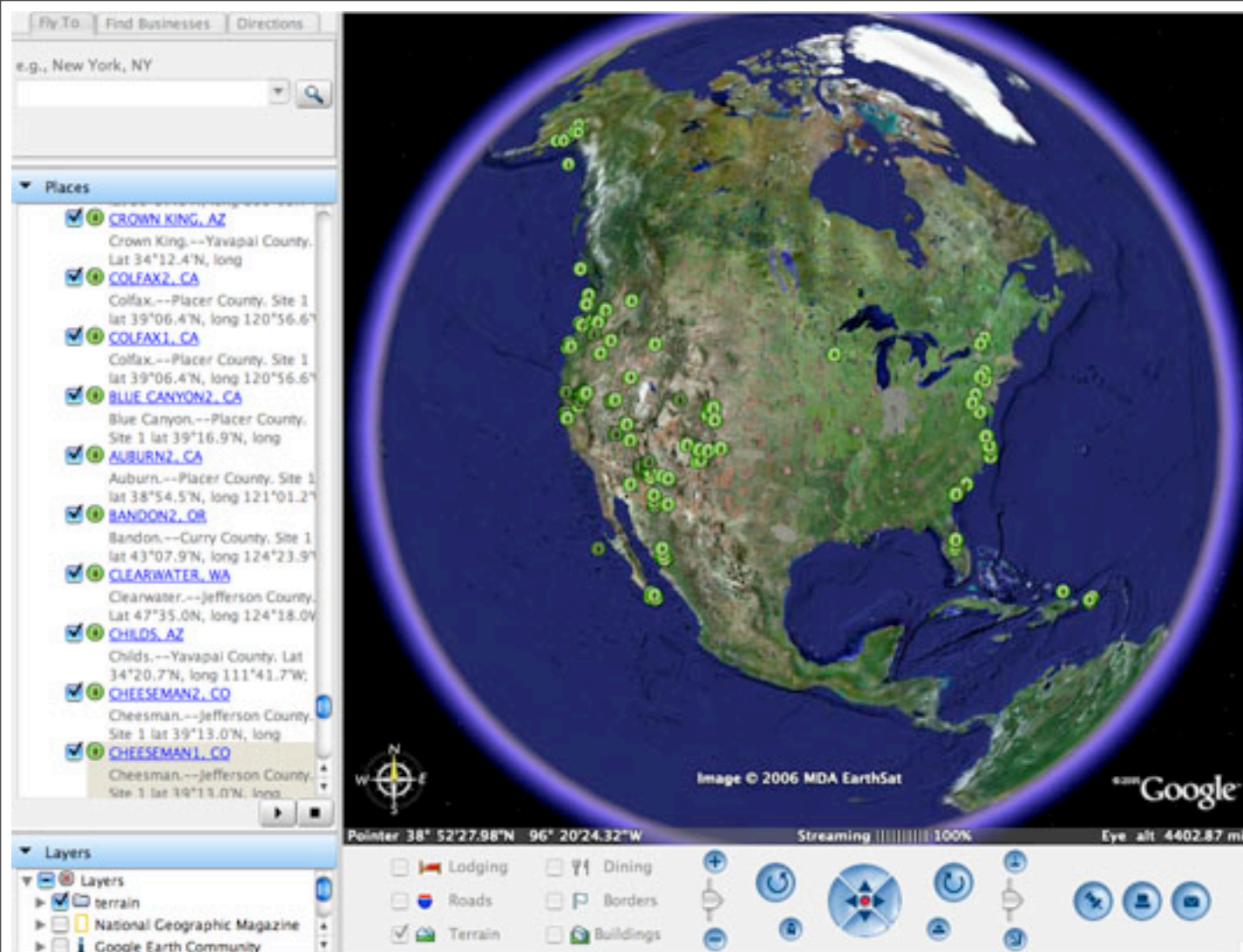
The numerical description is made up of 31 numbers representing an overall percentage score for each of the 31 characters states.

For the PHYSG3BR calibration set the result is a two dimensional array of 144 x 31 numbers.

◇	A	B	C	D	E	F	G	H	I	J	K	L	M	N	O	P	Q
1	Sample	Lobed	No Teeth	Regular te	Close teeth	Round tee	Acute teet	Compound	Nanophyll	Leptophyll	Leptophyll	Microphyll	Microphyll	Microphyll	Mesophyll	Mesophyll	Meso
2	Guanica, Puerto Rico	3	83	6	9	9	8	6	1	6	17	30	24	14	3	1	
3	Cabo Rojo, Puerto Rico	0	87	7	7	13	0	2	5	10	13	25	27	15	4	1	
4	Mocuzari A, Sonora	2	74	5	5	26	0	3	17	8	17	18	17	13	5	3	
5	Mocuzari-B, Sonora	3	74	5	6	24	2	0	12	13	22	25	17	8	1	1	
6	Natua, Fiji	0	79	15	9	18	4	2	0	0	1	8	21	21	23	15	
7	Borinquen, Puerto Rico	0	83	7	6	15	2	4	1	5	21	30	35	14	1	1	
8	Cambalache, Puerto Rico	0	87	10	4	12	1	3	3	5	7	16	28	23	12	3	
9	Tres Hermanos, Sonora	6	75	4	6	23	2	2	11	8	21	24	17	13	4	2	
10	Keka, Fiji	0	88	5	4	8	5	0	0	0	0	12	25	31	20	5	
11	Guajatica, Puerto Rico	1	72	14	8	19	9	4	1	2	3	10	19	26	21	10	
12	Susua Alta, Puerto Rico	0	80	4	4	20	0	4	4	12	20	22	18	11	5	4	
13	Cabo San Lucas, Baja California S	6	82	7	8	18	3	5	23	24	22	19	8	3	1	0	
14	Quiriego, Sonora	4	71	9	13	23	5	4	17	9	15	17	19	11	7	4	
15	Seqaqa, Fiji	0	61	16	11	38	2	7	0	0	0	8	16	17	17	26	
16	Nuri, Sonora	5	62	9	12	33	5	6	16	8	12	20	24	16	5	0	
17	Santiago, Baja California Sur	7	81	9	7	14	4	1	15	15	23	21	15	6	4	1	
18	Alamos, Sonora	5	66	8	9	28	5	4	9	8	13	18	21	19	8	3	
19	Empalme, Sonora	0	77	10	14	17	6	6	24	24	26	16	9	2	0	0	
20	Baie d'Magenta, New Caledonia	6	69	6	3	27	4	0	0	0	6	26	35	18	8	3	
21	Avon Park, Florida	5	73	16	13	13	15	2	2	5	9	26	32	16	8	2	
22	Orlando, Florida	7	68	16	17	28	4	3	2	5	14	17	33	22	6	1	
23	Todos Santos, Baja California Sur	4	81	5	7	16	2	1	25	27	28	10	8	3	1	1	
24	Buena Vista, Puerto Rico	0	81	10	6	18	1	3	0	6	14	25	30	15	7	1	
25	San Bartolo, Baja California Sur	3	69	15	19	26	5	3	20	16	20	16	16	10	3	0	
26	Canyon Lake, Arizona	2	78	7	10	18	3	0	41	29	16	9	6	0	0	0	
27	Los Divisaderos, Baja California Su	0	78	4	2	16	4	2	12	18	18	22	14	9	5	2	
28	Maricao, Puerto Rico	1	81	9	4	19	0	0	0	0	5	17	29	26	13	5	
29	Riv. Bleue, New Caledonia	0	57	10	9	35	9	4	0	0	0	8	12	14	24	22	
30	Bartlett Resvr., Arizona	0	80	10	13	10	10	5	30	20	24	18	6	2	0	0	
31	Mt. Koghis, New Caledonia	0	69	19	12	26	5	4	0	0	2	7	15	19	23	18	
32	Lake George, Florida	13	52	23	13	40	8	4	0	2	9	23	29	23	12	1	
33	Castle Cr., Arizona	2	73	10	14	24	3	2	26	26	19	16	9	3	3	0	
34	Silver Bell, Arizona	7	80	13	13	11	9	6	43	23	20	7	6	0	0	0	
35	Saguaro Lake, Arizona	4	75	13	15	22	3	2	21	28	22	17	12	1	0	0	
36	Superior, Arizona	2	69	16	16	26	5	5	35	20	18	17	9	2	0	0	
37	Roosevelt Lk., Arizona	0	72	9	14	25	3	2	24	30	22	13	8	2	1	0	
38	Point Mable, Georgia	12	63	10	18	24	13	52	2	4	8	24	33	18	5	5	
136	Akagawa Spa, Chima	11	79	64	56	25	3	39	0	0	5	16	23	24	19	7	
137	Republic, Washington	30	20	62	64	39	41	45	0	0	10	25	40	19	3	1	
138	Wanakana, New York	15	23	67	65	20	57	50	0	1	12	26	44	9	5	3	
139	Hanawa-Obono, Honshu	13	19	60	56	31	50	29	0	0	0	5	30	33	22	6	
140	Teshio, Hokkaido	12	15	65	58	31	54	40	0	0	0	2	25	24	23	15	
141	Kogawa, Hokkaido	14	19	64	61	22	59	44	0	0	2	6	30	30	18	10	
142	Tadenoumi, Honshu	18	7	82	73	20	73	61	0	0	0	9	31	31	21	7	
143	Lake Placid, New York	23	10	71	68	25	65	58	0	7	10	25	29	20	5	3	
144	Suganuma, Honshu	12	0	94	91	18	82	71	0	0	0	12	39	30	14	3	
145	Nukabira, Hokkaido	21	9	78	69	26	66	48	0	1	2	2	18	40	25	8	
146																	

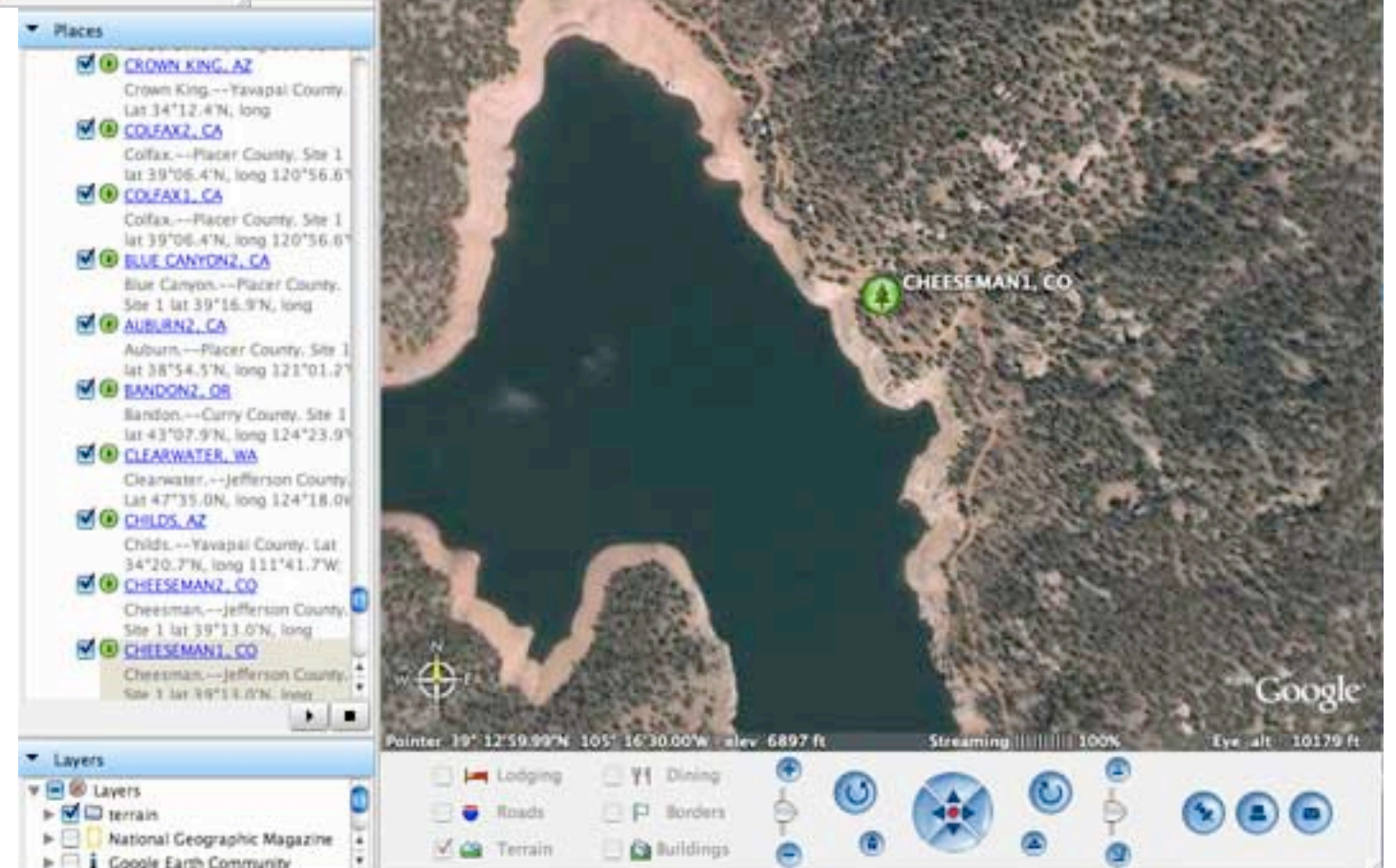
This percentage score is then added to a data array of modern calibration sites similarly scored. This is an example of the Physg3br calibration dataset. An unknown site for which a climate prediction is required is added to the end of such an array.





The geographical locations of the existing CLAMP calibration sites can be seen using Google Earth after downloading a .kmz file from the CLAMP website.

Most of the existing sites represent temperate vegetation in N.America, Europe and Japan.

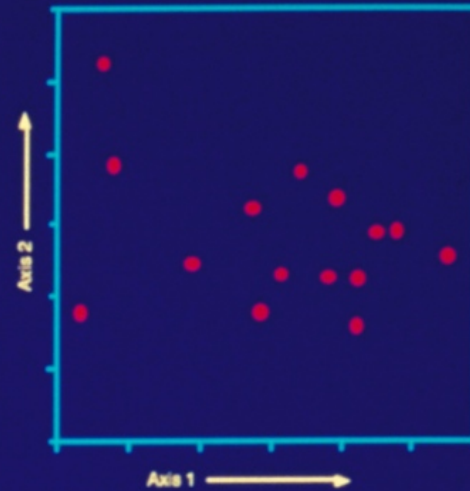






## Correspondence Analysis

**Taxon by  
Locality  
Data  
Matrix**



CLAMP uses a multivariate statistical engine called Correspondence Analysis (CA).

Unlike other multivariate methods such as multiple regression CA does not assume that the variables are independent.

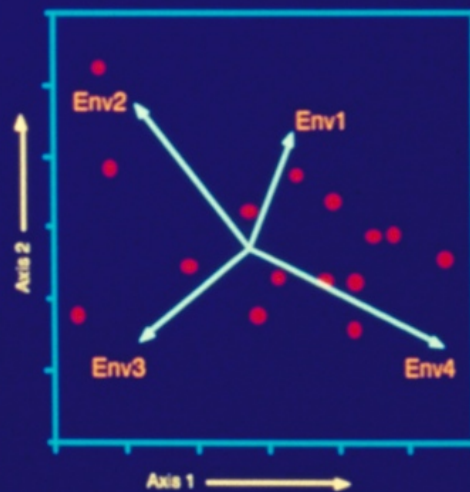
CA is also robust to missing data and is less sensitive to some variables having rare values.

In a variant called Canonical Correspondence Analysis a second data array consisting of environmental data is used to position environmental vectors in the ordination.

This is what is done in the current version of CLAMP.

## Canonical Correspondence Analysis

**Taxon by  
Locality  
Data  
Matrix**



**Environmental  
Variable by  
Locality  
Data  
Matrix**





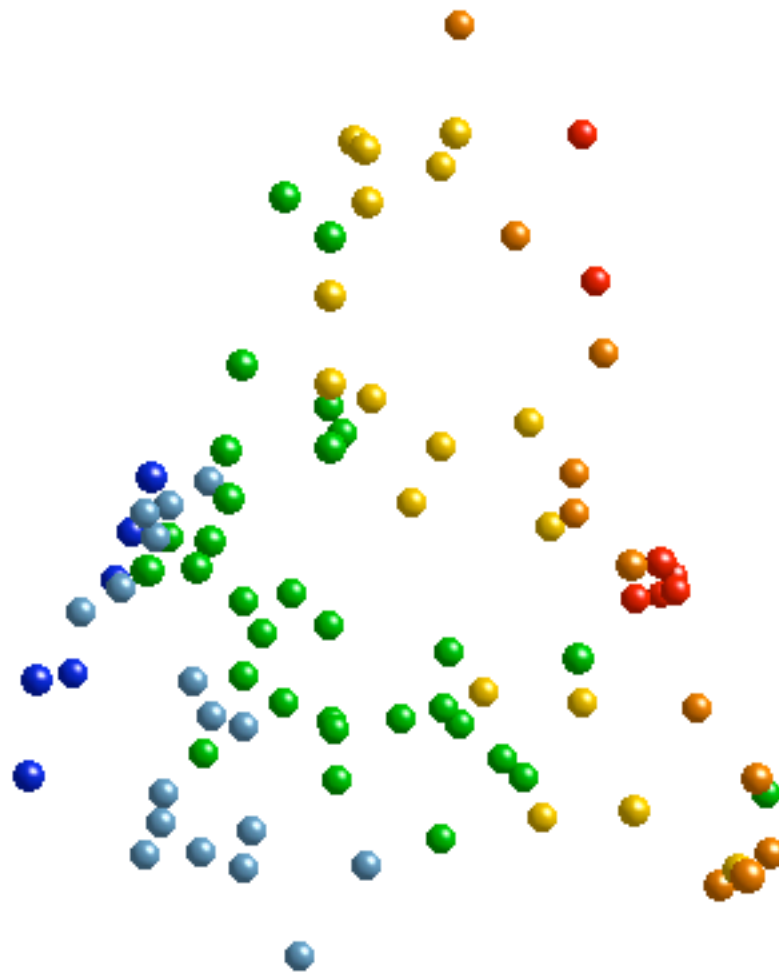
Part of the  
MET3BR  
modern  
observed  
meteorological  
data array.

MET3BR.xls													
	A	B	C	D	E	F	G	H	I	J	K	L	M
1	Sample	MAT	WMMT	CMMT	GROWSE	GSP	MMGSP	3-WET	3-DRY	RH	SH	ENTHAL	
2	Guanica, Puerto Rico	26.8	28.2	25	12	66	5.5	32	5.7	66	14.4	33.85	
3	Cabo Rojo, Puerto Rico	26.8	28.2	25	12	73	6.1	32	7.8	70	15.3	34.06	
4	Mocuzari A, Sonora	25.8	32.1	19	12	70	5.8	45	2.3	50	10.5	32.51	
5	Mocuzari-B, Sonora	25.8	32.1	19	12	70	5.8	45	2.3	50	10.4	32.48	
6	Natua, Fiji	25.6	26.7	24.3	12	290	24.2	105	22	74	15	33.86	
7	Borinquen, Puerto Rico	25.5	27.7	24.6	12	82	6.8	36	7.5	70	14.4	33.73	
8	Cambalache, Puerto Rico	25.5	26.9	23.5	12	154	12.8	48	29	79	15.9	34.06	
9	Tres Hermanos, Sonora	25.3	31.5	17.8	12	63	5.3	42	1.7	53	10.6	32.47	
10	Keka, Fiji	25.2	26.7	23.6	12	251	20.9	100	33	75	14.8	33.77	
11	Guajatica, Puerto Rico	24.8	26.3	23.2	12	190	15.8	61	27	82	15.6	33.85	
12	Susua Alta, Puerto Rico	24.5	25.6	22	12	188	15.7	72	19.8	71	14.4	33.68	
13	Cabo San Lucas, Baja California S	24.4	29.1	20.1	12	26	2.2	17.7	0.2	55	10.3	32.88	
14	Quiriego, Sonora	24.2	31.2	16.5	12	70	5.8	47	1.9	51	10.2	32.32	
15	Seqaqa, Fiji	24	25.5	22.4	12	280	23.3	115	23	84	15.5	33.82	
16	Nuri, Sonora	24	31.7	16	12	71	5.9	43	6.1	50	9.8	32.18	
17	Santiago, Baja California Sur	23.6	30.4	16.8	12	36	3.0	24	0.4	57	10.3	32.79	
18	Alamos, Sonora	23.5	30.4	16.2	12	77	6.4	47	2.3	54	10.1	32.17	
19	Empalme, Sonora	23.3	29.1	17.1	12	18	1.5	14	0.5	68	11.9	32.53	
20	Baie d'Magenta, New Caledonia	23	26.1	19.9	12	101	8.4	35	16	71	13.1	33.12	
21	Avon Park, Florida	22.4	27.7	15.6	12	134	11.2	60	14	72	12.2	32.88	
22	Orlando, Florida	22.2	27.9	15.3	12	122	10.2	55	14	70	11.7	32.71	
23	Todos Santos, Baja California Sur	22.1	27.7	18.4	12	18	1.5	11.2	0.4	65	10.8	32.77	
24	Buena Vista, Puerto Rico	22	23.8	21.1	12	181	15.1	77	20	83	14.7	33.44	
25	San Bartolo, Baja California Sur	22	28.1	16.5	12	36	3.0	26	0.6	58	9.9	32.53	
26	Canyon Lake, Arizona	21.9	33	11.6	12	35	2.9	11.8	2.4	41	5.6	31.13	
27	Los Divisaderos, Baja California Su	21.8	26.7	17.1	12	46	3.8	32	0.5	55	9.4	32.4	
28	Maricao, Puerto Rico	21.7	22.9	20.2	12	231	19.3	99	24	78	14	33.22	
29	Riv. Bleue, New Caledonia	21.5	24.7	18.2	12	230	19.2	93	19	79	12.6	32.83	
30	Bartlett Resvr., Arizona	21.4	32.4	11.5	12	35	2.9	10.9	2.7	40	5.3	30.99	
31	Mt. Koghis, New Caledonia	21	24.1	17.9	12	174	14.5	81	21.1	80	12.1	32.6	
32	Lake George, Florida	21	27.2	13.7	12	141	11.8	66	18	73	11.2	32.49	
33	Castle Cr., Arizona	20.9	31.6	11.5	12	40	3.3	12	2.5	34	4.8	30.86	
34	Silver Bell, Arizona	20.8	30.3	11.3	12	31	2.6	16	1.5	41	5.1	30.92	
35	Saguaro Lake, Arizona	20.6	31.7	10.7	12	33	2.8	10	2.2	41	6.1	31.13	
36	Superior, Arizona	20.4	30.3	10.9	12	47	3.9	15.9	3.6	43	4.5	30.71	
37	Roosevelt Lk., Arizona	19.8	31.5	8.7	11.2	32	2.9	11.8	2.8	42	5.5	30.9	
38	Brunswick, Georgia	19.6	27.8	11.1	12	134	11.2	55	21	70	9.8	31.97	
39	Anbo-west, Yakushima	19.2	27	11.1	12	429	35.8	159	75	75	9.6	31.64	
40	Nagakubo, Yakushima	19.2	27	11.1	12	429	35.8	159	75	75	9.6	31.64	
41	Monte Guilarte, Puerto Rico	19	20.5	17	12	223	18.6	81	35	86	13.5	32.82	
42	Beaufort, South Carolina	19	27.4	10.2	12	127	10.6	48	19	69	9.1	31.69	
43	Punkin Center, Arizona	18	29.8	7.6	9.6	35	3.6	12	2.8	42	5.5	30.8	
44	Yakusugi 260 m, Yakushima	17.9	25.7	9.8	11.5	418	36.3	158	36	75	9.5	31.39	
45	Toro Negro, Puerto Rico	17.9	19.7	16.6	12	238	19.8	101	26	86	13.1	32.6	
46	Childs, Arizona	17.9	29.3	7.7	9.7	33	3.4	16.2	4.3	39	5.3	30.69	
47	Simmons ville, South Carolina	17.8	26.9	8.5	10	114	11.4	48	24	66	8.4	31.46	





#### Vegetation Site MAT



Here is a CANOCO plot of the Physg3br modern calibration sites, colour coded for the Mean Annual Temperature (MAT) under which they were growing. The relative positions of the sites in physiognomic space are determined by leaf architecture.

Here is a CANOCO plot of the Physg3br modern calibration sites, colour coded for the Mean Annual Temperature (MAT) under which they were growing (using the observations in the Met3br file). The relative positions of the sites in physiognomic space is determined by leaf architecture alone. The Met3br file only calibrates the structure of physiognomic space in terms of climate. Note that although this plot is two dimensional if static or three dimensional is you are viewing this as a movie in reality the technique analyses the site data in 31 dimensional space defined by the leaf character state scores. Almost 70% of all the structure in 31 dimensional space resides in the first three dimensions so this way of looking at physiognomic space is a reasonable representation of reality.



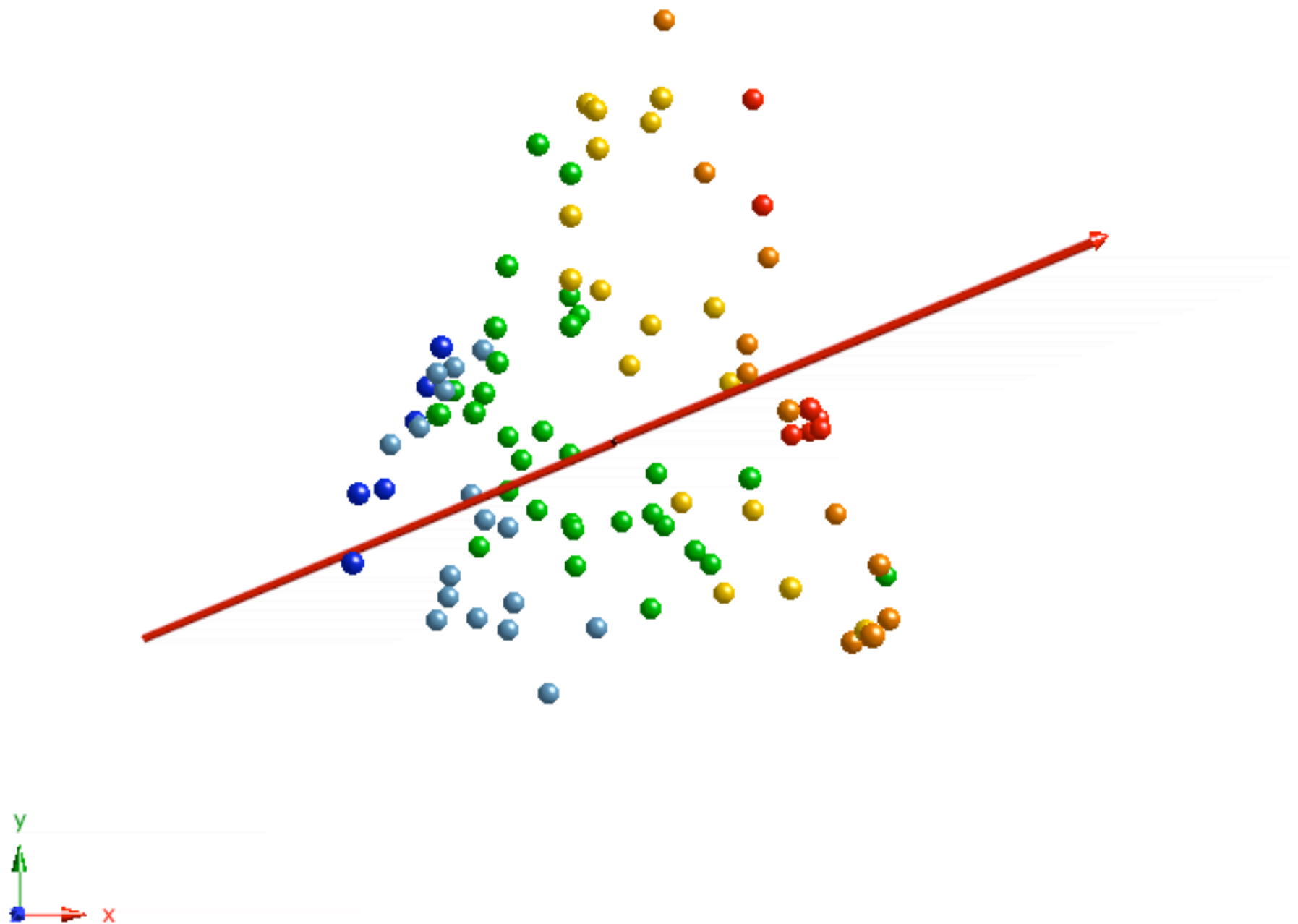


# Vegetation Site MAT

- > 25°C
- 20 - 24.9°C
- 15 - 19.9°C
- 10 - 14.9°C
- 5 - 9.9°C
- < 5°C

## Climate Vectors

→ MAT



There is a clear trend from cool to warm sites that can be summarized by an MAT vector.



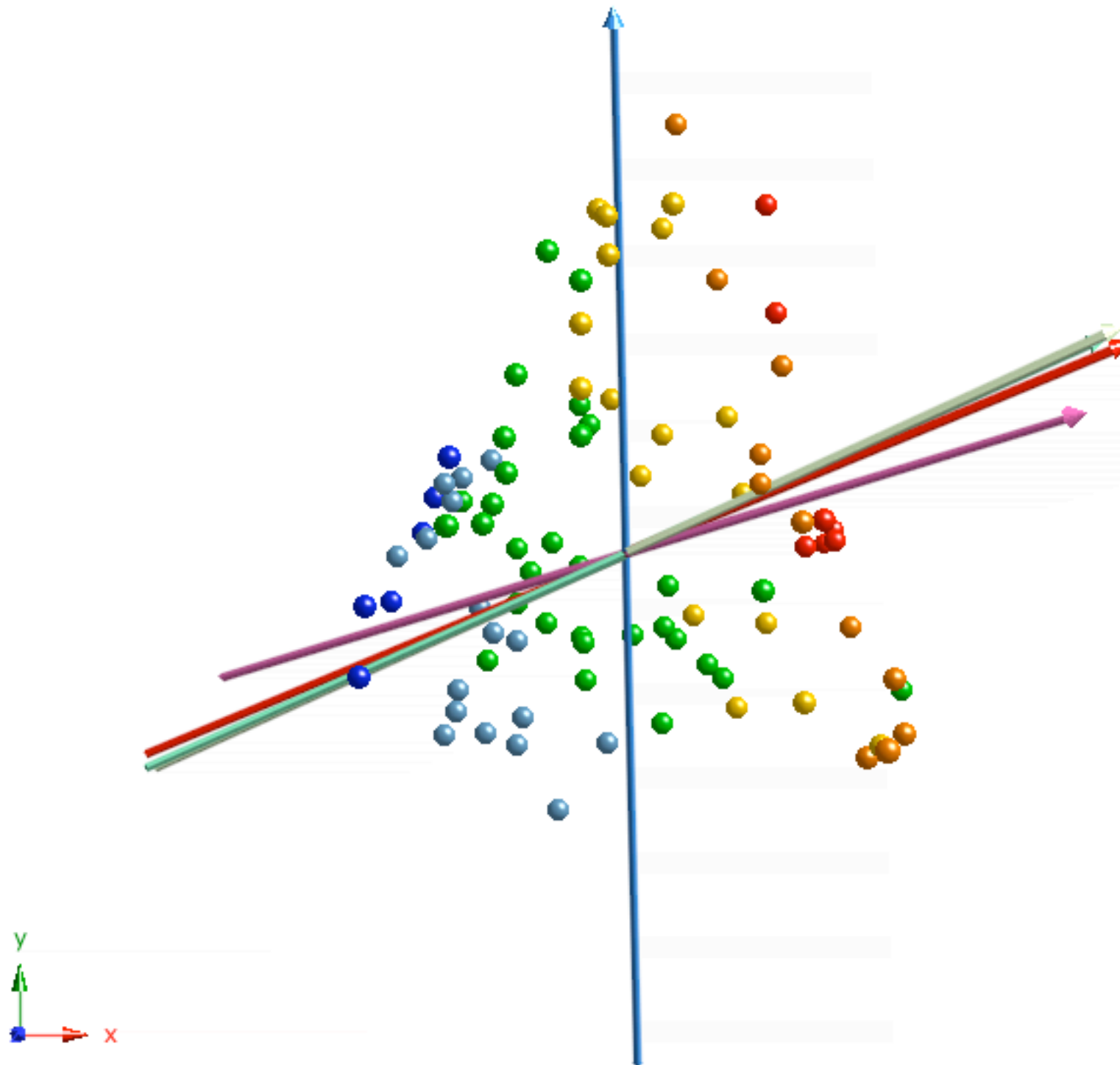


#### Vegetation Site MAT

- > 25°C
- 20 - 24.9°C
- 15 - 19.9°C
- 10 - 14.9°C
- 5 - 9.9°C
- < 5°C

#### Climate Vectors

- MAT
- WMMT
- CMMT
- GROWSEAS
- GSP



Other vectors for warm month mean temperature (WMMT), cold month mean temperature (CMMT), length of the growing season (GROWSEAS) and growing season precipitation (GSP) can also be added.



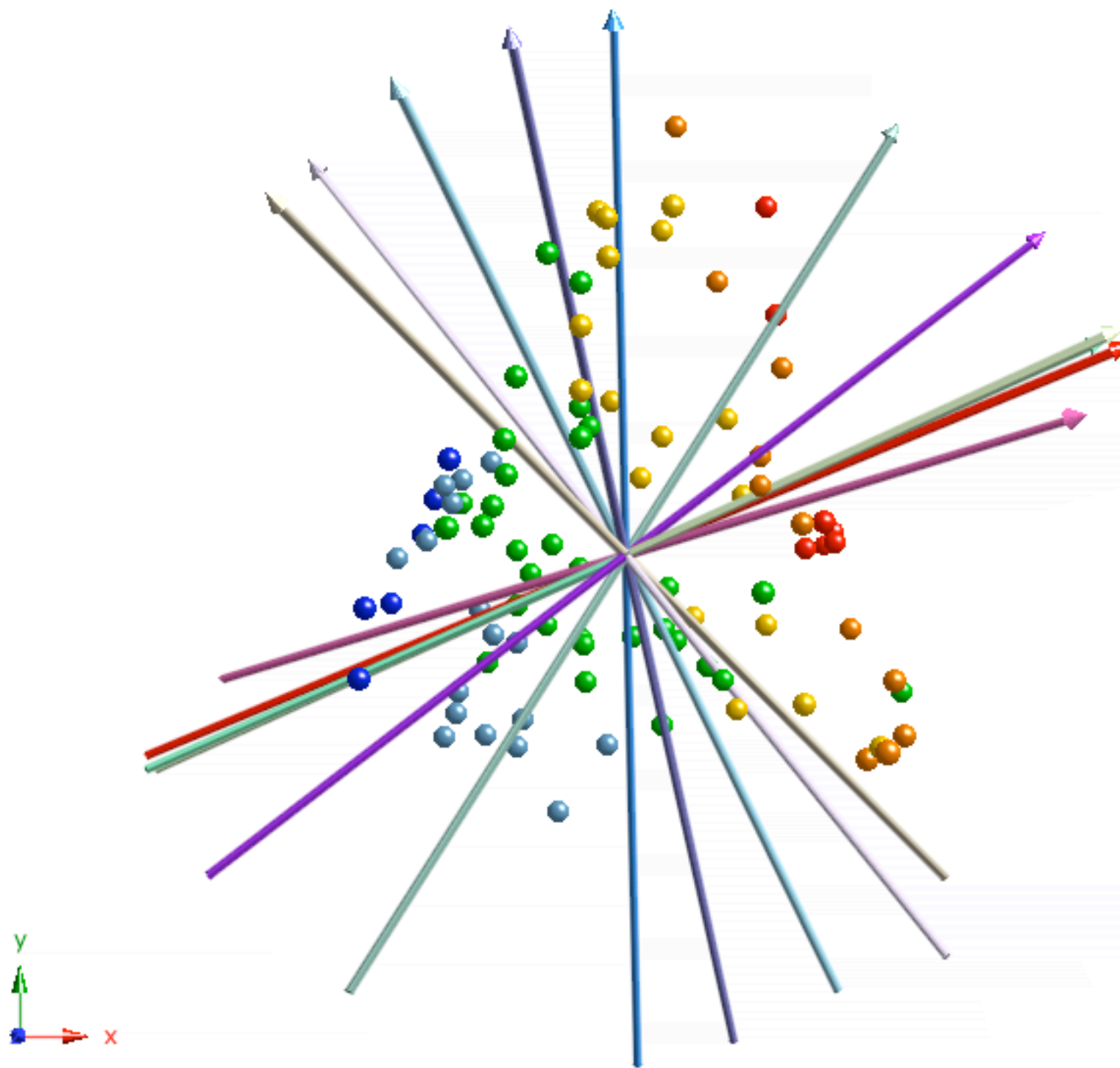


#### Vegetation Site MAT

- > 25°C
- 20 - 24.9°C
- 15 - 19.9°C
- 10 - 14.9°C
- 5 - 9.9°C
- < 5°C

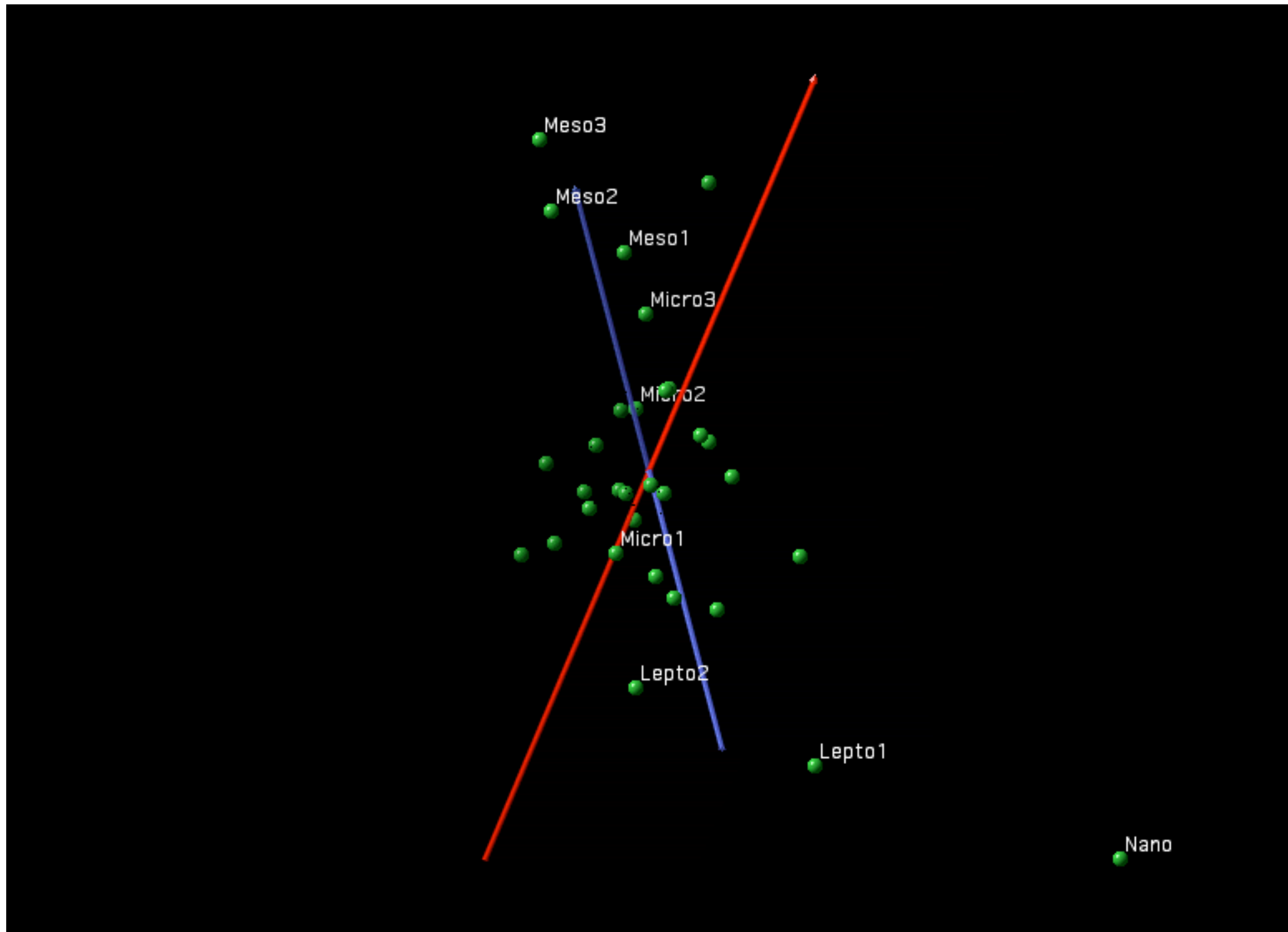
#### Climate Vectors

- MAT
- WMMT
- CMMT
- GROWSEAS
- GSP
- MMGSP
- 3-DRY
- 3-WET
- RH
- SH
- ENTHALPY



All the CLAMP vectors can be calibrated because the climates under which the modern sites are growing are known.





As well as the vegetation sites being plotted it is also possible to plot the positions of the leaf characters in physiognomic space. This shows the relationships between the characters and the MAT and GSP vectors.

As well as the vegetation sites being plotted it is also possible to plot the positions of the leaf characters in physiognomic space. This shows the relationships between the characters and the MAT and GSP vectors. One of the characteristics of the CANOCO statistical technique is that this plot is directly comparable to the site plot but here 31 character states are plotted in a 3-D representation of 144 dimensional space (the number of sites in the Physg3br data array. For clarity only the size categories have been labelled and in this case increasing size describes a helical shape in 3-D, the long axis of which is aligned roughly with the blue precipitation vector. Note that alignment is weak suggesting factors other than water availability are related to leaf size.

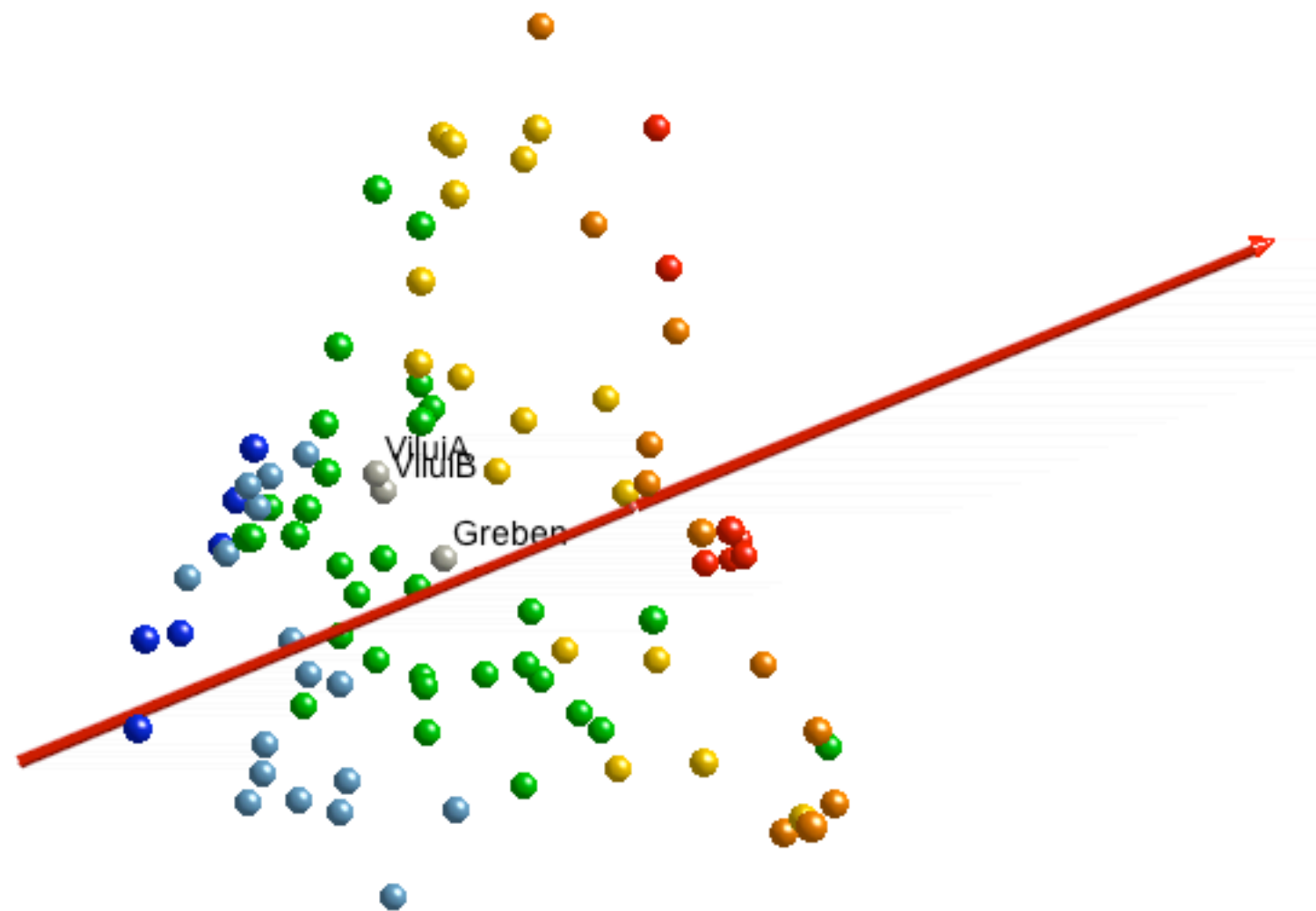




#### Vegetation Site MAT



#### Climate Vectors

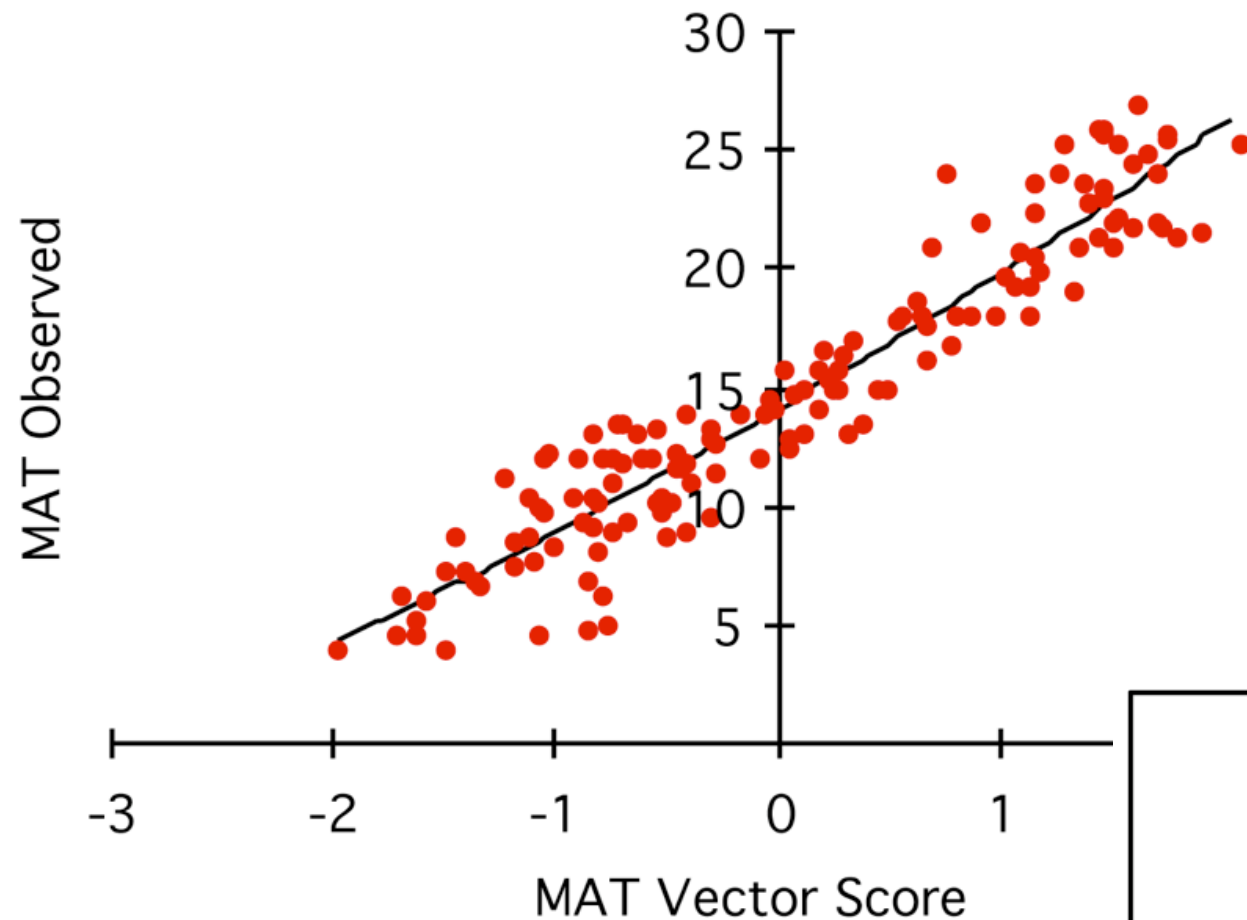


The grey points labelled “Greben”, “ViluiA” and “ViluiB” represent fossil samples (in this case from the Cretaceous of Russia). Their positions along the calibrated vectors yield estimates of ancient climate.





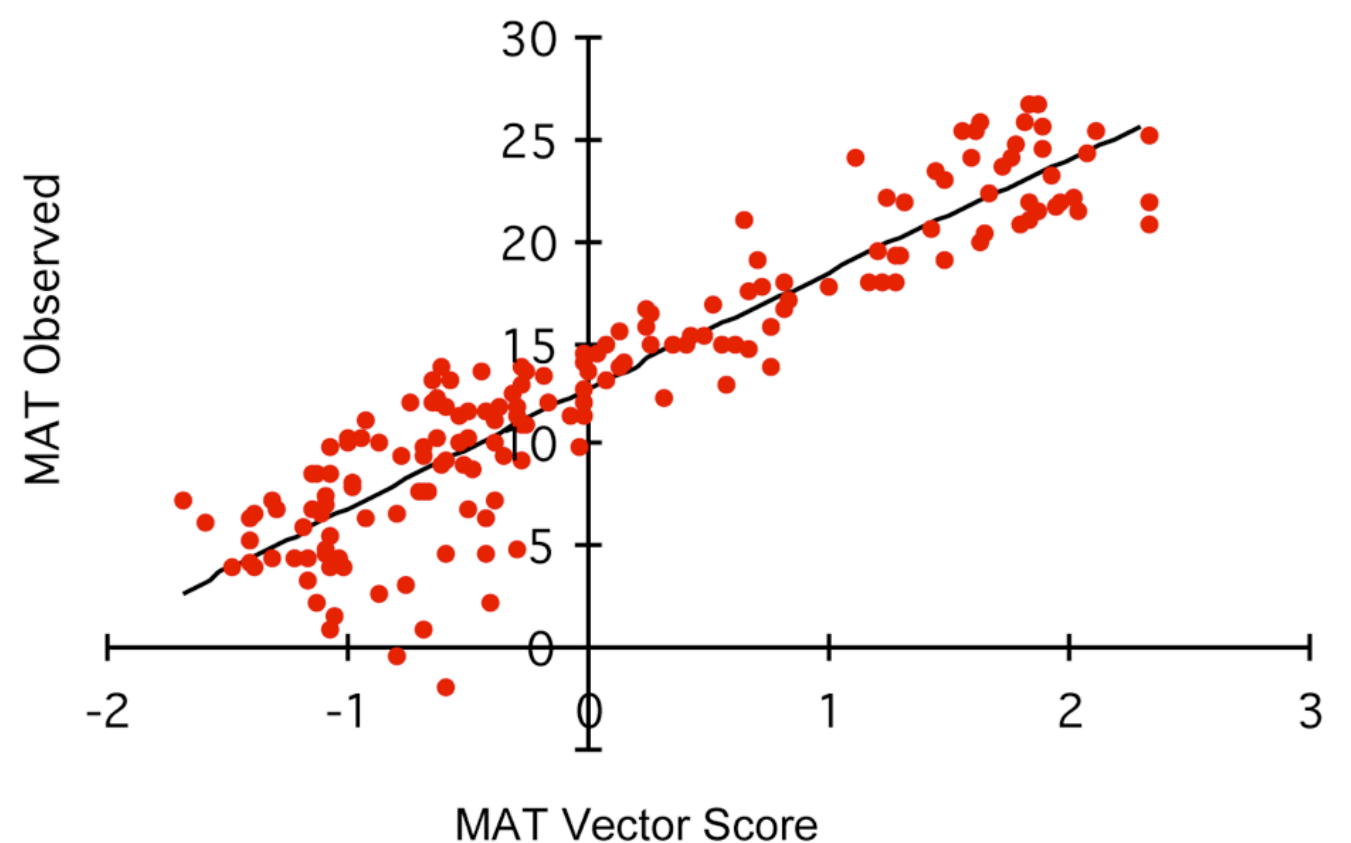
### Mean Annual Temperature



This graph shows the observed Mean Annual Temperature (MAT) plotted against the MAT vector score for the Physg3br calibration data set.

This graph shows the observed Mean Annual Temperature plotted against the MAT vector score for the Physg3ar calibration data set.

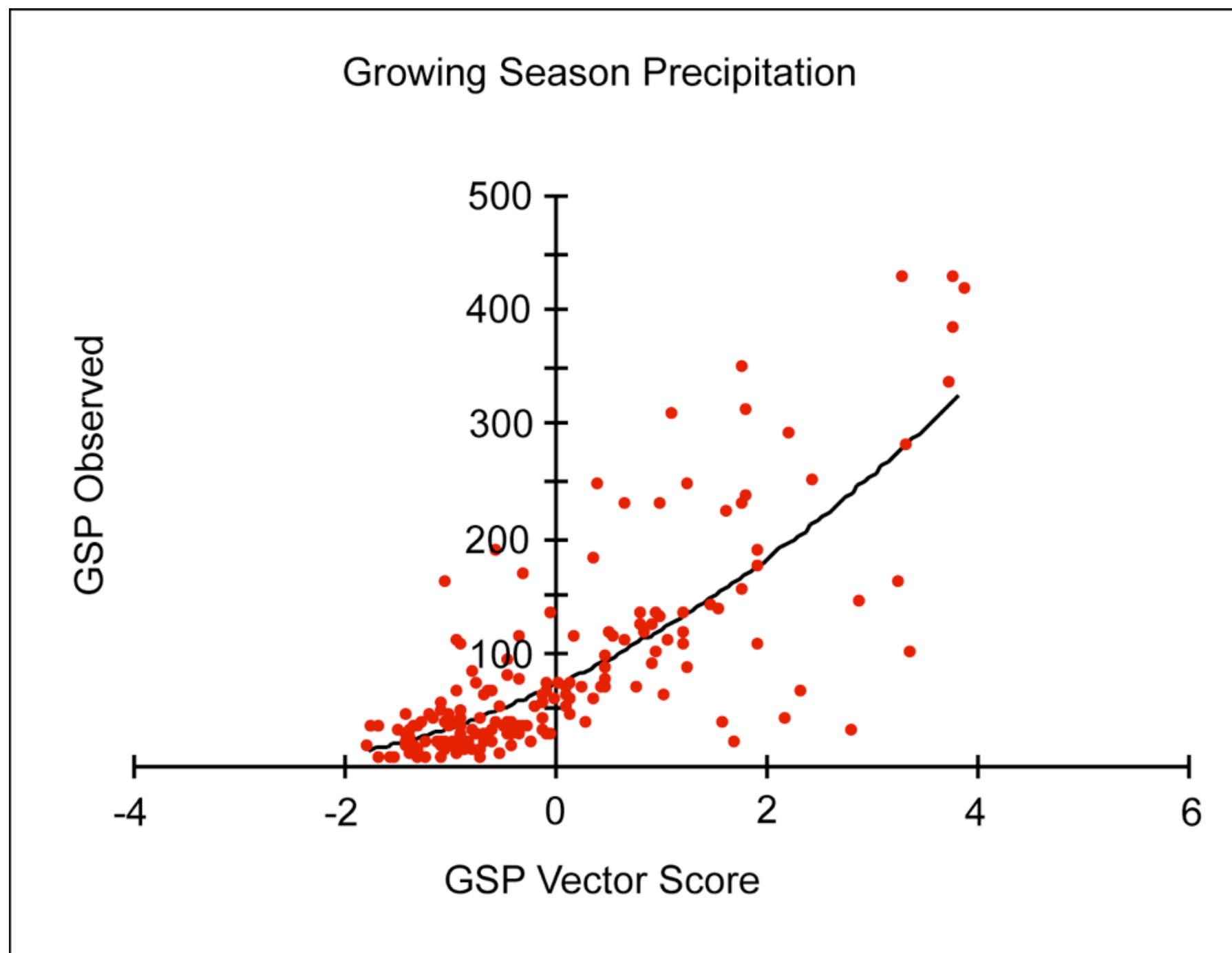
### Mean Annual Temperature



The upper graph shows the observed Mean Annual Temperature (MAT) plotted against the MAT vector score for the Physg3br calibration data set. The spread of the points about the regression line (a 2nd order polynomial) provides a measure of the statistical uncertainty of the technique. These uncertainties, the regression equations and the graphs are given in the spreadsheets (denoted Res3br for the Physg3br and Met3br data sets and downloadable from the CLAMP website) used to calculate the climate predictions.

The lower graph shows the observed Mean Annual Temperature plotted against the MAT vector score for the Physg3ar calibration data set. This data set includes sites where freezing conditions are significant. In general this data set should only be used where cold conditions are suspected because the errors are larger than with the Phys3br data set.



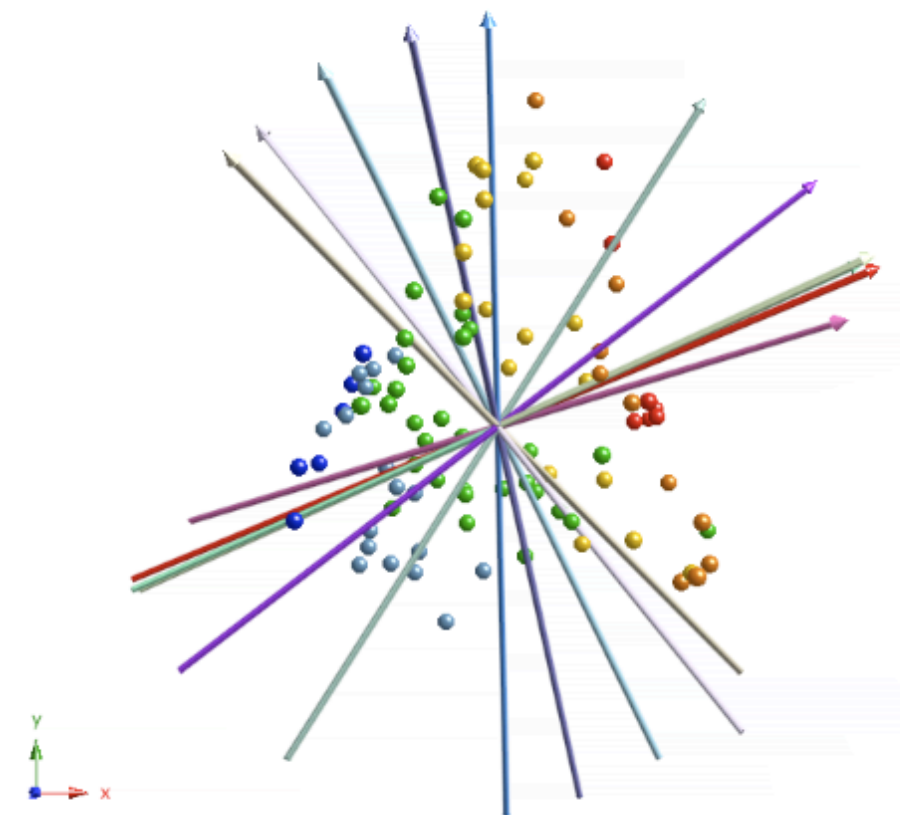


Observed growing season precipitation plotted against the GSP vector score. In dry environments leaf physiognomy maps on to precipitation better than in wet environments.





CLAMP offers the most diverse insights into past climates and inevitably is continually being refined and improved. New developments include the use of globally gridded meteorological data and, in due course, the vector based approach will be replaced by the use of nearest neighbours (*sensu* Stranks and England, 1997, The use of a resemblance function in the measurement of climatic parameters from the physiognomy of woody dicotyledons. *Palaeogeography, Palaeoclimatology, Palaeoecology*, 131: 15-28).



For the latest information visit the CLAMP website:

<http://tabitha.open.ac.uk/spicer/CLAMP/Clampset1.html>