

CLIMATE RECONSTRUCTION FOR THE LATE PLIOCENE – EARLY PLEISTOCENE

WITH QUANTITATIVE PALEO-CLIMATE METHODS FOR THE NETHERLANDS

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Climate reconstruction for the Late Pliocene and Early Pleistocene – with quantitative paleo-climate methods for the Netherlands

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Preface and Acknowledge

This guided research about quantitative paleo-climatic reconstruction for the Late Pliocene and Early Pleistocene for the Netherlands has been written as part of the Master Science degree in Earth Science, programme Marine Science at Utrecht University. The research was performed under the supervision of Dr. Timme Donders and Dr. Thomas Giesecke. The aim of this guided research is to reconstruct the climate during the Late Pliocene – Early Pleistocene for the Netherlands based on pollen assemblage data with quantitative paleo-climatic reconstruction methods. Three different fossil data (Noordwijk, Petten and Hank) sets are used to make the reconstruction, whereby the dataset from Whitmore is used as reference data set. The analysing program C2 is used to reconstruct the winter temperature and total precipitation. The technique used for this guided research are transfer functions using Modern Analogue Technique, Weighted Average and Weighted Average Partial Least Squares algorithms. It was a very interesting subject for me because I was unfamiliar with this way of reconstruction. I have learned a lot about reconstruction methods and the way to use them.

First, I would like to thank my supervisor Timme Donders for his great effort and enthusiasm to complete my guided research. I could always text him or we had interesting teams meetings to explore the analysing program and to discuss the data. It was hard for me to work every day on the same desk in the same room as I lived because of the Covid-19 pandemic, however, due to the effort of my friends, roommates, and parents I finished this guided research on time.

Abstract

This guided research aims to reconstruct the climate during the Late Pliocene / Early Pleistocene for the Netherlands based on pollen assemblage data with quantitative paleoclimatic reconstruction methods. The Late Pliocene and Early Pleistocene are important Epochs because of their characteristics by cyclic growth and decay of terrestrial ice sheets related to the Milankovitch cycles. The Late Pliocene through Early Pleistocene is a frequently targeted interval for paleo-environmental reconstructions because it is considered as an analogue for future climate change. The focus of this guided research is on the Netherlands. Three different terrestrial sections are used from different parts of the Netherlands: Noordwijk, Petten and Hank. The Noordwijk section is from the south-western part of the Netherlands and has a time span from 2.61 - 2.11 Ma years ago, Early Pleistocene. The Petten section has an age range of 1.35 - 2.41 Ma years ago, Early Pleistocene and the Hank section originates from the Late Pliocene (5.45 - 2.48 Ma years ago).

The modern pollen-based quantitative paleo-climatology has undergone the development of a diverse array of statistical techniques to transform fossil assemblage data into past climate estimates. Three main approaches for quantitative reconstruction of past climates from biostratigraphical data are known: indicator species approach, assemblage approach and multivariate transfer function approach. The multi-variate calibration function approach involves underlying statistical models with global estimation of parametric functions for all the taxa present. The analysing program C2 is used to reconstruct the winter temperature and total precipitation. Pollen-climate transfer functions were calculated using WA with inverse deshrinking and 3- component WA-PLS. Calibration set species data values were square-root transformed for WA and WA-PLS regression to reduce noise in the data. Performance statistics were computed for each transfer function, these include the coefficient of determination (R^2) , Root Mean Square Error of Predication (RMSEP) and maximum bias. Both reconstructions are compared with the percentage deciduous trees in the samples and known data from Dearing Crampton-Flood et al., (2020) for the Late Pliocene and Zagwijn (1992) for Early Pleistocene. From these results, WA seems to fit the best in Late Pliocene data and WA-PLS fits best in the younger data.

WA seems to fit better in the Late Pliocene and WA-PLS fits better for the Early Pleistocene. The problem for WA-PLS is the that it tries to fit all the variation in the reconstruction, for older data this is not the right way because not all pollen taxa do exist today in the study area. Some mistakes inevitable if WA-PLS is used for older data than the Pleistocene. Further research requires more datapoints to make a better connection with other reconstructions.

Contents

Pre	Preface and Acknowledge						
Ab	stract	t	4				
1.	Intr	oduction	7				
2.	Gen	eral background information	8				
2	2.1	Different quantitative reconstruction techniques	8				
	2.1.2	1 Weighted Average (WA) and Weighted Average Partial Least Squares (WA-PLS)	9				
	2.1.2	2 Modern Analogue Technique (MAT) 1	0				
	2.1.3	3 Principal Component Analysis and Bootstrapping1	2				
2	2.2	Modern dataset 1	3				
3.	Mat	terial and methods1	4				
3	8.1	Location of the modern and fossil datasets1	4				
	3.1.2	1 Location Noordwijk 1	4				
	3.1.2	2 Location Petten 1	4				
	3.1.3	3 Location Hank 1	4				
3	8.2	Creating datasets and models in C2 1	5				
4.	Res	ults 1	8				
4	.1	General overview of the data 1	8				
	4.1.2	1 Ordination results 1	8				
4	.2 Cli	mate reconstruction	1				
4	.3	Optimum in the winter temperature and total precipitation 2	2				
	4.3.2	1 Late Pliocene reconstruction 2	2				
	4.3.2	2 Early Pleistocene reconstruction	3				
5.	Disc	cussion	6				
5	5.1	How accurate are the climate reconstructions?	6				
	5.1.2	1 Comparison with percentage deciduous trees 2	6				
	5.1.2	2 Comparison with other datasets 2	7				
	5.1.3	3 The differences between WA and WA-PLS 2	8				
5	5.2	How accurate is the data from Whitmore et al., (2015)?	9				
6.	Con	clusions	1				
7.	Refe	erences	2				
Lis	t witł	h figures	7				
Ap	pendi	ices	8				
A	A. Po	ollen diagram modern data set Florida3	9				

B.	Pollen diagram fossil dataset Noordwijk	. 45
C.	Pollen diagram Petten	. 48
D.	Pollen diagram Hank	. 51
E.	Dominant pollen species for fossil data sets with age	. 53
F.	Principal Component Analyse Modern and Fossil data	. 56
G.	Correspondence graph with all the data	. 57
H.	Observed graphs WA and WA-PLS	. 58
I.	Observed graphs WA and WAPLS with removed samples	. 62
J.	First climate reconstruction	. 66
K.	Calibration Sample number – age	. 67
L.	Fossil pollen species combined with modern pollen species	. 68
М.	Percentage Trees & Bisaccates	. 70

1. Introduction

Paleo-ecological and paleo-climatic research is done on almost every region on the world and have produced many fossil pollen records. The fossil pollen records are used to reconstruct past climates for many places on Earth. Most of the paleo-climatic reconstructions are done to reconstruct future climate changes and to indicate potential climate changes in the future.

The aim of this guided research is to reconstruct the climate during the Late Pliocene and Early Pleistocene for the Netherlands based on pollen assemblage data with a quantitative paleoclimatic reconstruction method. The sub-questions for the guided research are:

- Which quantitative paleo-climate technique is the best to reconstruct the climate for the Netherlands during the Late Pliocene and Early Pleistocene?
- Is the dataset of Whitmore et al., (2005) a good dataset to use for a climate reconstruction for the Late Pliocene and Early Pleistocene for the Netherlands?
- Is the summer temperature, reconstructed by Zagwijn (1992) for the Netherlands, in agreement with the reconstructed climate from our reconstruction method?

Three new data-sets with known ages are used for the reconstruction: Hank (Dearing Crampton-Flood et al., 2020) for the Late Pliocene and Noordwijk (Noorbergen et al., 2015) and Petten (Houben, 2019; Krom, 2020) for the Early Pleistocene. The techniques used for this research are transfer functions using Modern Analogue Techniques (MAT), Weighted Average (WA) and Weighted Average Partial Least Squares (WA-PLS) algorithms. MAT, WA and WA-PLS are statistical methods applied on a pollen assemblage data set from the south-eastern part of the United States with 67 different pollen taxa. The climate of the United States have the same climatic characteristics as during the Pliocene and Pleistocene of the Netherlands. The reconstructed summer temperature from the Netherlands for the Early Pleistocene originates from Zagwijn (1992). Zagwijn (1992) made a pollen based July temperature curve with a temperature range between 0 and 20 °C. The temperatures are inferred for the Reuverian (3.1 – 2.5 million years ago (Ma)) and are based on semi-quantitative indicator-taxa approach.

Zagwijn (1992) and Zagwijn (1996) described in his papers about the rarity of the use of terrestrial material instead of sea surface temperatures (SST), most reconstructions are largely based on SST (Dowsett et al., 2010). Most reconstructions are based on the Co-existence Approach (CA) which relies on the presence or absence of a plant taxon within a fossil assemblages. The two techniques WA and WA-PLS are techniques never used for the time interval Late Pliocene and Early Pleistocene, the techniques are based on a dataset from the USA to reconstruct Late Pliocene and Early Pleistocene data from the Netherlands.

2. General background information

The Pliocene (5.33 – 2.58 Ma) is an Epoch from the Neogene Period (23.02 – 2.58 Ma) and the Pleistocene (2.58 Ma – 11.7 thousands (ka)) and Epoch from the Quaternary (2.58 to present). Lisiecki and Raymo (2005) described the characteristics of the Late Pliocene and Early Pleistocene, "one of the characteristics is the cyclic growth of terrestrial ice sheets related to the Milankovitch cycles". The large terrestrial ice sheets started to develop in the Northern Hemisphere about 2.57 Ma years ago, resulting in multiple glacial – interglacial oscillation driven by variations in orbital insolation as a result of periodic fluctuations in Earth's orbit on Milankovitch timescales of 100 (orbital eccentricity), 41 (Earth's obliquity) and 19 – 23 (precession) ka intervals (Imbrie and Imbrie, 1980; Ruddiman, 1990, 2013; Bradley, 2015). Before about 1.25 Ma, glacial-interglacial oscillations appear to be symmetric with small ice volumes and periodicity of 41 ka years (e.g. Birks, 2019).

The most recent reconstruction for the Pliocene is made by Dearing Crampton-Flood et al., (2020), the reconstruction is based on material from the Southern North Sea Basin from the Hank location. Dearing Crampton-Flood et al., (2020) indicates a 2 to 6 °C warmer SST than present, this makes the Late Pliocene a good interval for prediction of future climate change. During the Late Pliocene and Early Pleistocene the position of the shoreline of the North Sea was positioned more inland with respect to the present-day coastline of the Netherlands (e.g. Westerhoff, 2009; Noorbergen et al., 2015). Ziegler (1990) described "*the Pliocene North Sea had only an opening to the Atlantic in the north, other sides were confined by several landmasses*". Some other connections were confirmed by Gibbard and Lewin (2016), however, only during periods of high sea level. The Southern North Sea basin has a good record of SST evolution and climate change for the north-western European continent because of the high terrestrial input by large European rivers: the Eridanos River and Rhine-Meuse river, described by Gibbard (1988) and confirmed by Dearing Crampton-Flood et al., (2020).

The fossil pollen records are used to reconstruct different climate variables, these reconstructions gave understanding of local climate change over time. Chevalier et al., (2020) described two kinds of dispersal strategies, whereby an-emophilous pollen depends on the wind. The wind mixes the grains in the air from various locations and spread them over large distances. The an-emophilous pollen grains are then further transported by the large European rivers and deposited in the Southern North Sea. So it is possible that not every pollen grain we found in our fossil data-sets originates from that specific location. We found possibly a mix of locally and remotely produced pollen grains in the pollen assemblages, whereby the locally pollen have a higher concentration. The remotely produced pollen have influence on the climate reconstructions and can indicate wrong climate characteristics if they are found in higher concentrations.

2.1 Different quantitative reconstruction techniques

Since the start of the 19th century researchers tried to reconstruct the climate of the Late Quaternary. Birks et al., (2010) described the early studies of Blytt (1881), Andersson (1909), Iversen (1944) and Von Post (1946). They used plant macro fossils, animal remains, pollen assemblages and peat stratigraphy to indicate the climate.

Three main approaches for quantitative reconstruction of past climates from bio-stratigraphical data are known (Birks & Birks, 1980; Birks, 1981; 1995; 1998; 2003). These techniques are:

• Indicator species approach

The indicator species approach indicates the presence or absence of one taxa. This technique is the first reconstruction method developed for climate reconstruction. Andersson (1903, 1909, 1910) used this technique to indicate the warm season by the presence of *Corylus avellana*.

• Assemblage or analogue approach The assemblage or analogue approach indicates the similarity measure between assemblages in all the samples. It considers the fossil assemblages as a whole and the relative abundance of all the different fossil taxa (e.g. Birks & Birks, 1980).

• **Multi-variate calibration function approach** The multi-variate calibration function approach involving calibration functions, it uses underlying statistical models with global estimation of parametric functions for all the taxa present (e.g. Birks et al., (2010)).

For this research different multi-variate calibration functions are used, MAT, PCA, WA and WA-PLS. MAT is a calibration method to reconstruct a past environmental parameter, it works by finding modern sites with faunal associations close to the ones in the samples. PCA is used to describe a huge amount of data in smaller components, the principal components. WA and WA-PLS assume a unimodal organism-environmental response model. The difference between both models is the dimensionality, WA uses full dimensionality and WA-PLS a reduced dimensionality.

The co-existence approach, developed by Mosbrugger and Utescher (1997), is the most used method for quantitative paleoclimatic reconstructions. The technique is based on the 'nearest living relative philosophy', the technique corresponds to the 'indicator species approach' of Birks and Birks (1980). The co-existence approach is based on two taxa with a fossil relevance, the best estimate of the mean annual temperature under which the fossil taxa once lived, is the interval where they both lived (Mosbrugger and Utescher, 1997). The co-existence approach is very sensitive and focuses on the extremes in the climate, but it does not show the details. This is a disadvantage, the technique does not use the relative abundance of a fossil taxa, it only uses the presence or absence. For example, the research of Prader et al., (2020), little variation in the temperature is observed while the quantity of specific pollen taxa are constantly changing. The relative abundance gives additional information to the model, but it is not included.

Mosbrugger and Utescher (1997) did some research about the use of the CA approach, the best results are given for the Oligocene and younger because most of the nearest living relatives are still present. For the Palaeocene and Eocene floras the results are at a lower climatic resolution because the nearest living relatives can only be identified on family level (Mosbrugger and Utescher, 1997). For older material it is difficult to use the technique because there are problems with identifying nearest living relatives.

2.1.1 Weighted Average (WA) and Weighted Average Partial Least Squares (WA-PLS)

The assumption of WA and WA-PLS is 'a plant species have their highest abundances in their own climate niches'. The most abundant types are closest to their optimal climatic conditions at the time the pollen assemblages are deposited. The model estimates the climate optima from a calibration set, in this case the dataset from Whitmore et al., (2015), specified on SE USA. Birks et al., (2010) describes the basic idea of WA as "*if a taxon shows an unimodal relationship with a*"

particular climate variable x, its abundance will tend to be highest at sites with values of x close to the taxon's environmental optimum or niche peak". The taxon's optimum for x is the average of all the x values for every site, divided by the taxon's relative abundance.

WA and WA-PLS were first used to reconstruct water chemistry variables in the palaeolimnology (e.g Chevalier et al., 2020). WA has some disadvantages, the modern sample are most of the time unevenly distributed in relation to the environmental gradient of interest, estimation of the optima may be biased (e.g. ter Braak and Looman, 1986). Another weakness is the 'edge-effect', whereby low variables are overestimated, and high values are under-estimated (e.g. Birks et al., 2010; Braak & Juggins, 1993). A solution to reduce the edge-effect is to apply a correction of inferred values referred to as de-shrinking (Birks, 1995). There is estimated how much the range is reduced by modelling a regression model between the reconstructed and the target values from the modern samples. The reconstructions are then rescaled with respect to the range of the target (Nolan et al. 2019).

Ter Braak et al., (1993) developed an improvement on the WA based on partial least squared regression, WA-PLS. WA-PLS is developed to address some of the problems identified in WA, it combines the advantages of WA with the multi-collinearity reduction approach of PLS. WA-PLS suffers less from the edge-effect than WA. It is important that not too many components are selected because this will give the model a poor predictive power. In general, WA-PLS models based on two or three components perform best.

Both methods have been used widely to reconstruct e.g. Late Pleistocene and Holocene precipitation in China (Chen et al., 2015; Li et al., 2017; Liet al., 2012; Lu et al., 2018) and temperature in Siberia (Klemm et al., 2013) and Europe (Finsinger et al., 2010; Lotter et al., 2012; Lotter et al., 2000; Sakari Salonen et al., 2013; Seppä et al., 2009; Veski et al., 2015). WA-PLS performs well in various ecological and climatic conditions (Juggins and Birks, 2012). The models are less sensitive to the size of the studied climate gradient (Chevalier et al., 2020). However, some pollen taxa are multimodal for the variable of interest, mostly by continental or global studies.

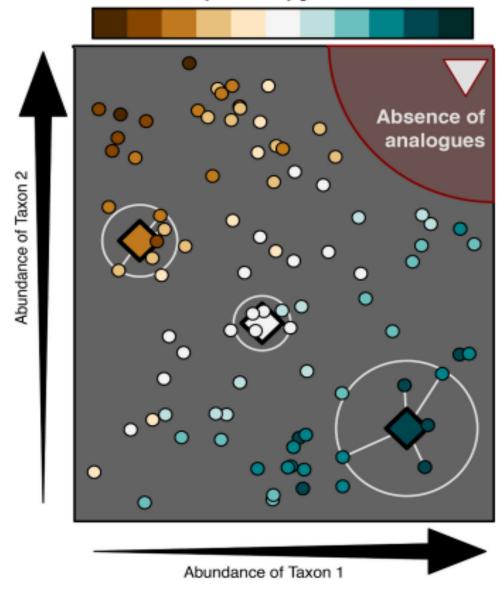
2.1.2 Modern Analogue Technique (MAT)

MAT is a statistical technique called 'k-nearest-neighbours'. The technique is based on the measure of degree of similarity between fossil pollen assemblages and modern assemblages (Jackson and Williams, 2004). The most similar modern samples, with weights being the inverse of the dissimilarities so that modern samples with the lowest dissimilarity have the greatest weight in the climate reconstruction (e.g. Birks et al., 2010; Overpeck et al., 1985). "*Having found the modern sample(s) that is (are) most similar to the fossil assemblage, the past climate for the fossil sample is inferred to be equivalent to the state of the climate variable(s) of interest for the analogues modern sample(s)"* (Birks et al., 2010).

"A 'no-analogue' situation occurs when none of the assemblages from the training set resemble the fossil assemblage with sufficient accuracy to be considered an acceptable analogue and can be caused by either a restricted collection of modern samples or by past conditions that have no counterparts in the modern settings" (Chevalier et al., 2020). Williams and Shuman (2008) did a study about broad-scale cross validation across North-America and comparing modern samples against themselves. They concluded that a small number of closest analogues, from 3 to 7, outperforms the use of only the closest analogue (Williams and Shuman 2008). Including more analogues increases the risk of false positive matches, especially when the calibration dataset encompasses wide spatial areas where the low taxonomic resolution of pollen data can lead to similar pollen compositions to be observed despite experiencing very different climates (Chevalier et al., 2020). William and Shuman (2008) did research about this topic and have

proposed that several broadscale North American taxa, like *Pinus, Alnus* or *Tsuga*, could be split into eastern and western pollen taxa to circumvent this issue. When focussing on regional splitting of taxa, a better climatic response is estimated.

The figure below indicates a good example of MAT and is made by Chevalier et al., (2020). The figure is an example of the aridity-humidity gradient, whereby taxon 1 (green) prefers humid conditions and taxon 2 (brown) prefers drier conditions. Three fossil pollen assemblages (diamonds with thick black borders) are compared with the modern assemblages, each had its own humidity value, the coloured circles in the figure. The five closest modern assemblages (white circles) are averaged and estimated the past conditions.



Aridity – Humidity gradient

Figure 1 Conceptual diagram describing MAT. Taxon 1 prefers humid climates (green) and taxon 2 prefers drier conditions (brown). Three fossil pollen assemblages (diamonds) are compared to a modern set, each one associated with a humidity value (coloured circles). The five closest analogues (white circle) are retained and averaged to estimate past conditions. (Chevalier et al., 2020)

2.1.3 Principal Component Analysis and Bootstrapping

PCA is a multi-variate analyse method to describe a huge amount of data into a smaller one that still contains most of the information in the large dataset. The purpose of the technique is to find a new variable in the dataset with a maximal variance. Normally, reducing the number of variables of a dataset decreases the accuracy of a dataset, however, it makes the data simpler. The variables in the dataset are called the 'principal components' and indicate the maximum variation in the data. This technique makes it easier to analyse the dataset due to the smaller datasets that are created with the same amount of information.

The figure below is made by Chevalier et al., (2020) and represents a conceptual diagram of WA-PLS. Three different taxa are used, the black part indicates all the taxa and are not considered in this example. The graph on the right indicates the ecological gradient of the taxon, the modern assemblages. Climate values are extracted from each modern sample, one taxon occurs in and weighted by the corresponding pollen percentages to derive the climate optimum of the taxon. the small circles on the right side. For more information about WA and WA-PLS, there is referred to the paper of Chevalier et al., (2020).

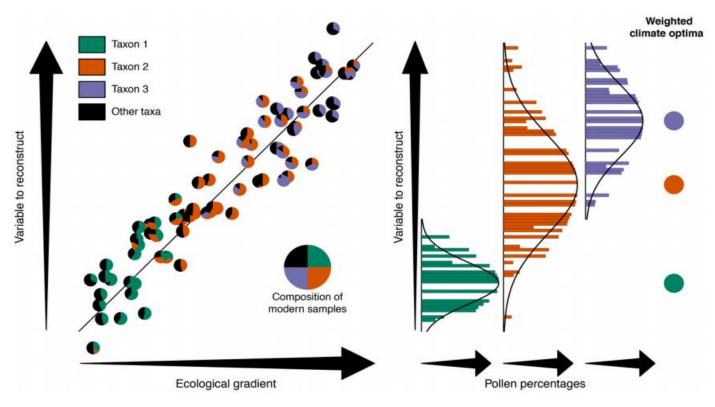


Figure 2 Conceptual diagram describing WA-PLS. The black part indicates all the taxa which are not considered in the example. On the right, the ecological gradient is showed, the modern assemblage. Climate values are extracted from each modern sample to derive the climate optimum of the taxa, the small circles on the right (Chevalier et al., 2020).

Bootstrapping is used as an approach for each model. Chevalier et al., (202) explained bootstrapping as "*if we call n the number of samples composing the modern data set, the bootstrap approach consists of randomly sampling n samples with replacement to create a new dataset of the same size*". The unselected samples become part of the test data set, however, some samples are selected multiple times randomly. For each model run, a specific bootstrap set is chosen and used for the specified reconstruction.

2.2 Modern dataset

The dataset contain 4634 sites from across North America and contain four types of information for each record: (1) pollen counts for 134 pollen taxa that are either found across North America or are regionally important, (2) site identification, geographic coordinates , source, depositional environment and auxiliary identification codes, (3) environmental data, including elevation-corrected climatic data based on the Climatic Research Unit gridded climatology (New et al., 2002) with improved lapse rate correction and (4) information about natural and potential vegetation derived from cartographic sources (Whitmore et al., 2005).

Most of the pollen data are from lake sediments (2261 sites), however mostly in the southeastern part of the USA, samples are taken from moss pollsters (860 sites). Most of the lake sediments are from Canada, while samples from the south-eastern US, northern Pacific Coast and the Central Plains are from moss pollsters and terrestrial/soil samples from the southwestern part of the US. The original data for the modern pollen database came from coretop samples originating from many sites where fossil data were analysed during the past 40 years and local and regional surface sample datasets accumulated expressly for calibration studies (Whitmore et al., 2005). The pollen are processed following a standard procedure with a small amount of sediment and some series of acid and bases to remove the sedimentary material, the pollen grains remain intact (e.g. Whitmore et al., 2005). The classification of the pollen is based on taxonomic and morphological hierarchies defined in the NAPD (Grimm, 2000a).

A three-step procedure was used to indicate the climate and bio-climate variables. First, the local lapse rate for the long-term monthly mean temperature, precipitation and possible sunshine data were calculated (e.g Whitmore et al., 2005). Second, the lapse rate at the surrounding 10-min grid points were used to adjust the sunshine data values at those points to the elevation of the target point (e.g. Whitmore et al., 2005). Third, the values were interpolated at the location of the target point using geographically weighted bilinear interpolation (e.g. Whitmore et al., 2005). Whitmore et al., (2005) defined the bioclimatic variables as *"calculation from the basic climatic variables including growing-degree days with a 0 and 5 °C, a chilling period, mean temperature of the coldest and warmest month of the year and the January/annual and July/annual precipitation ratios".*

3. Material and methods

3.1 Location of the modern and fossil datasets

Three different fossil datasets from the Netherlands: Noordwijk, Petten and Hank are used for the climate reconstruction. Svenning (2003) described Pliocene and Pleistocene extinctions in Europe and the correspondence in cool-temperate tree genera for North-America and Europe. Some species occurring during Late Pliocene and Early Pleistocene in Europe, are still present in North America (Florida region) and are now extinct in Europe (e.g. Svenning, 2003). The modern dataset from Whitmore et al., (2015) originates from North America and Greenland and is limited to the SE USA region for this guided research. The Florida region is chosen because of the climate characteristics, the climate conditions during the Late Pliocene and Early Pleistocene for the Netherlands were almost the same as the current climate conditions in Florida. The winter temperature is used for the reconstruction because the summer temperatures in Florida are likely higher comparison with the climate conditions in Netherlands during the Late Pliocene and Early Pleistocene and Early Pleistocene. Also, winter temperatures are more limiting to warm temperate to subtropical vegetation, so it is more likely to reflect realistic trends by using summer temperatures.

3.1.1 Location Noordwijk

Noordwijk is in the southwestern part of the Netherlands (Figure 3) in the province of South-Holland. The environment around Noordwijk is created during one of the last ice periods due to the interaction of the sea and the Old Rhine river. The dataset of Noordwijk was drilled under authority of the Geological Survey of the Netherlands (Rijksgeologische Dienst: RGD) using air lifting drill methods (e.g. Noorbergen, 2015). The 73 samples were taken at a depth from 454.1 – 52.5 m and have an age from 2.61 – 2.11 Ma years ago and contain 45 different pollen types. The borehole consists of different Formations, mainly the lower Maassluis, Westkappele Ground and Upper Red Grag Formation. For more information about the Noordwijk core there is referred to Noorbergen (2015).

3.1.2 Location Petten

Petten is located at the North Sea in the northwestern part of the Netherlands (Figure 3) in the province North-Holland. Two different boreholes were drilled in Petten, borehole 1 (BH1) is used for the reconstruction. The top of the borehole is at an elevation of 3.26 m above Normaal Amsterdams Peil (NAP). Down to 101,50 m depth, sonic drilling was performed. Below that, wireline rotary coring was performed. TNO (2019) partially describes the core and said that the core was recovered in unconsolidated, dominantly sandy Pleistocene sequences. The borehole was taken at a depth from 383,60 – 196.85 m depth, has 71 samples and contain 33 different pollen species. The samples of Petten have an age range of 2.41 – 1.35 Ma. For more information about the Petten core there is referred to TNO (2019) and the thesis from Krom (2019).

3.1.3 Location Hank

Hank is in the southern part of the Netherlands (Figure 3), within the current Rhine-Meuse-Scheldt delta. The site is located within the Ruhr Valley Rift, a region that experienced relatively high tectonic subsidence during the Late Cenozoic (e.g. van Balen et al., 2000; Dearing Crampton-Flood et al., 2020). The borehole was drilled by air-lifting technology to a base of 404 m in 2001. The Geological Survey of the Netherlands (TNO) and Dearing Crampton-Flood et al., (2020) described the lithology of the core, the base consists of the upper part of the marine Breda Formation, followed by sandy and clay-rich marine belonging to the Oosterhout Formation. Sometimes the marine sediments contain shell fragments and crags. The dataset contain 27 samples and 43 different pollen species. The age ranges from 2.48 – 5.42 Ma years ago. For more information about the Hank core there is referred to Dearing Crampton-Flood et al., (2020).



Figure 3 Location of the fossil datasets, from north to south: Pettten, Noordwijk and Hank

3.2 Creating datasets and models in C2

The combination of the pollen species in the dataset from Whitmore et al., (2015) presents a specific climatic pattern, the characteristics of each sample are presented in the environmental datasheet. As described in paragraph 2.3, the south-eastern part of the United States is used for this guided research. The latitude range was the limited factor to choose the samples from the dataset of Whitmore et al., (2015). The latitude range is 25° - 50° N, the value is determined because of the sample amount and the pollen species occurring in between this region. The bimodal distribution of the pollen is important to keep in mind, mainly for *Pinus*. This genus has different species, occurring in warmer (lowland) and colder (mountains) environments. The gradient is used to include both *Pinus*, however, it could gave problems for other species like the Cupressaceae and Taxodiaceae group. The longitude ranges from -52.6° - 99.3° E.

The three fossil datasets contain pollen taxa which do not exist anymore or do not occur in the modern dataset because of climate limits. The modern and fossil dataset must be at the same taxonomical resolution otherwise the program cannot link them with each other. The merged dataset can be found in Appendix K, the fossil pollen species have been combined with other pollen species from the same family to make a new dataset. Alisma, Armeria, Azolla (Mosquito

Fern), *Brassicaceae.*, *Dipsacaceae*, *Engelhardia*, *Eucommia*, *Isoëtes*, *Lythrum*, *Myriophyllum*, *Parthenociccus*, *Pilularia*, *Platycarya*, *Potamogeton*, *Reveesia*, *Sapotaceae*, *Sciadopitys*, *Symplocos*, *Trudopollis*, *Typhaceae* and *Vitis* have no taxonomic representation in the modern pollen dataset and are not used for the reconstruction. *Carpinus* and *Ostrya* are merged into *Ostrycar*, the other fossil pollen species with another 'modern' name are mentioned in Table 1. The decision to call them by this specific family name usually stems from the family branch they origin from. Usually, the fossil pollen is a clade of the same family as the modern variant.

Fossil name	Family name	Merged with	
Asteraceaea	Asteraceae	Aster	
Calluna	Ericaceae	Ericaceaea	
Cedrus	Pinaceae	Larix	
С.	Asteraceae	Aster	
Tubuliflorae			
C. Liguliflorae	Asteraceae	Aster	
Dryopteris	Dryopteridaceae	Polypodiaceae	
Filipendula	Rosaceae	Rosaceae	
Gramineae	Poaceae	Poaceae	
Illex	Aquifoliaceae	Aquifoliaceae	
Juniperus	Cupressaceae	Cupressaceae	
Lonicera	Caprifoliaceae	Caprifoliaceae	
Nymphaea	Nymphaeaceae	Liliaceae	
Rhus	Anacardiaceae	Anacardiaceae	
Sequoia	Cupressaceae	Taxodium	
Spergula	Caryophyllaceae	Caryophyllaceae	
Succisa	Caprifoliaceae	Caprifoliaceae	
Thelypteris	Thelypteridaceae	Polypodiaceae	
Umbelliferaea	Apiaceae	Apiaceae	
Viscum	Santalaceae	Arceuthobium	

Table 1	Family	names a	of the	fossil	ones
Tuble 1	runniy	numes c	y une	103311	ones

The fossil datasets have been merged into one big dataset with the data from Noordwijk (1 - 73), Petten (74 - 145) and Hank (146 - 172). Appendix K shows the combination of the part of the dataset from Whitmore et al., (2015) and the fossil datasets. The first three columns show the pollen species from the fossil datasets and the fourth the names used in the dataset. Thereafter, a new pollen sum and pollen percentages are calculated.

Two multi-variate functions (PCA and Correspondence Analysis (CA)) are used to indicate the correspondence between the data from Whitmore et al., (2015) and the fossil data is. After analysing the first results, *Pinus*, and *Pinus haploxylon*-type are eliminated from the data because of the outside position in the graphs. *Pinus* has a bimodal distribution, it has two different peaks, one in the cold part of the gradient and one in the warm part of the gradient. The bi-modal distribution caused that *Pinus* is not a good pollen species to use as an indicator for climate reconstruction. The first PCA is done with all the data to confirm the outliers of the CA graph, the second one is done without the sample outliers and *Pinus* and *Pinus haploxylon*-type. Second, MAT is used to observe the reliability (test of 'closeness') of the fossil samples to the reference data, the values are used to describe the reliability of the reconstructions. The distance to the closest modern analogue is used to describe the reliability, the 'goodness of fit' measured by the 1st and 5th percentile with the reference sample, calculated as squared residual distance, are considered very good and good analogues, respectively (Jackson and Williams, 2004). One

percentile indicates that 99% of the data has a close analogue in the reference data. The percentiles are showed in Table 4. The last step is the calculation of the pollen-climate transfer function using WA and WA-PLS with inverse de-shrinking for WA and 3rd component for WA-PLS. Prentice (1980) suggest to square root transform species data for WA and WA-PLS regression to reduce noise in the data. Two parameters are used to reconstruct the climate: the winter temperature and the total precipitation. The summer climate in Florida has more tropical characteristics than during the Late Pliocene and Early Pleistocene in the Netherlands. The same reason could be given for the choice to reconstruct the total precipitation, during the tropical warm summers a lot of water evaporates. Also, Florida has a kind of monsoon period with a lot of precipitation, probably different in comparison with the Pliocene in Europe. The calculation for the mean winter temperature and total precipitation of WA and WA-PLS transfer function was performed with C2 version 1.7 software (Juggins, 2016) for the period 5.45 – 1.34 Ma years ago. Both methods were used to indicate which method is better to use for this time period. Most papers use the co-existence approach for this time period, this guided research will introduce transfer-functions like WA and WA-PLS to this time period. The performance statistics were computed for each transfer functions, these include the coefficient of determination (R^2) , R^2 bootstrapping and Root Mean Square Error or Prediction (RMSEP). These statistics are shown in Table 1 and Table 2.

For some samples, the age is interpolated because the age was unknown, the average of the sum of the sample above and below is calculated to determine the age. The reconstructed data is discussed, and the minima and maxima temperature and precipitation are explained by the combination of pollen found in the samples. The results from the WA and WA-PLS reconstruction are compared with each other to decide which method is better to use for this period. WA is the simplest approach with the least assumptions, it assumes an unimodal relationship between species and environments. This relationship is tested by plotting the gradient of species against the temperature (Appendix A). WA-PLS used residual variance to increase the model fit. Also, the percentage of deciduous trees is determined to indicate potential climate cycles.

4. Results

Appendix A, B, C and D show the pollen diagrams of Whitmore et al., (2015), Noordwijk, Petten and Hank, respectively. The dominant pollen taxa (Appendix E) in Noordwijk and Petten are *Alnus, Artemisia, Betula, Corylus, Cyperaceae, Ericaceae, Picea, Pinus, Poaceae, Polypodium, Quercus* and *Sphagnum*. For Petten *Alnus, Cheno/Am, Corylus, Ericaceae, Picea, Pinus,* Poaceae, *Polypodium, Quercus* and *Sphagnum*. The dataset of Hank has five dominant pollen species: *Abies, Picea, Pinus, Polypodium* and *Taxodium*. The three datasets have a dominance in the bisaccate pollen, *Pinus* is already removed from the figure. *Pinus* has also a high dominance in the modern dataset in comparison to other pollen species. One explanation for the high percentage of *Pinus* pollen is the easy way of dispersion of the pollen. Bisaccate pollen are easily transported by the wind and are most sensitive to differential transport processes (e.g. Donders et al., 2018). Because of this effect, it cannot be said with certainty of the *Pinus* pollen, found in the fossil and modern datasets, are regionally deposited.

4.1 General overview of the data

Appendix E shows the dominant pollen species with a timescale (2.56 – 2.12 Ma) on the y-axis for Noordwijk. The first observation is the occurrence of *Polypodium* until an age of 2.287 Ma years ago and the dominance of *Ericaceae* for the whole section. *Poaceae* has its highest dominance during 2.38 – 2.29 Ma. Another observation is the decrease of *Polypodium* and *Poaceae* at almost the same moment, also *Picea* has a remarkable decrease and do not fluctuate that much anymore. *Betula* has one peak at an age of 2.29 Ma. *Sphagnum* and *Alnus* fluctuate for the whole section between 5 and 20%. *Aster* has a peak at an age of 2.55 Ma.

The Petten dataset has an approximate time range from 2.41 – 1.35 Ma years ago. The same decrease in *Poaceae* and *Polypodium* is recognised at an age of 2.287 Ma years ago. *Polypodium* has a new high peak at an age of 1.47 Ma years ago. For Noordwijk, *Ericaceae* is present for the whole section, the section of Petten indicates the decrease in *Ericaceae* from 1.67 Ma years ago. *Quercus* becomes more important at an age of 1.97 Ma years ago.

The time range of the dataset from Hank is from 5.43 – 2.48 Ma years ago. Only four pollen species are dominant for this dataset, these pollen species are *Abies, Picea, Polypodium* and *Taxodium. Taxodium* is the most dominant pollen species in this dataset, during the highest dominance periods the other dominant pollen species are almost zero. *Polypodium* is dominant for the period 2.98 – 2.5 Ma years ago.

4.1.1 Ordination results

The idea of PCA is to reduce the number of variables of the dataset, the new eigenvectors preserving as much information as possible. Appendix A shows the PCA without the removal of *Pinus* and *Pinus Haplox*, these figures show that the PCA's have very long gradients. Assumed is the linear relationship between the environment and the in- or decrease of species (Appendix A). Figure 4 shows the PCA for the modern and fossil data without *Pinus* and *Pinus haploxlyon*-type. The first subdivision spread the data into *Cheno/Am*, *Polypodium*, *Ericaceae*, *Picea* and *Abies* and *Betula*, *Tsuga*, *Ulmus*, *Quercus* and *Poaceae*. The lower part is clearly divided into *Quercus* and *Betula*, while for the upper part the division is less clear, but it is made between *Cheno/Am* and *Picea*. The fossil datapoints are more centred in the middle and upper part, whereby the Petten samples are spread over the whole upper part and Hank and Noordwijk almost only in the middle. A lot of *Polypodium* and *Ericaceae* are observed in these samples.

PCA Modern + Fossil data

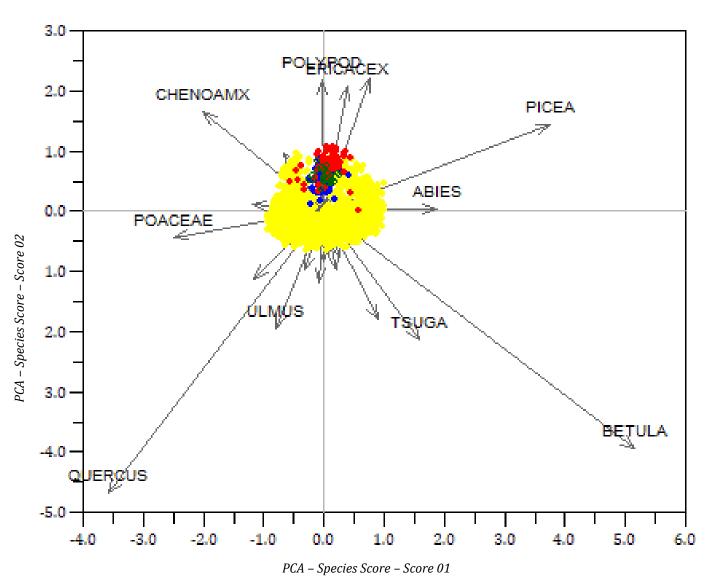
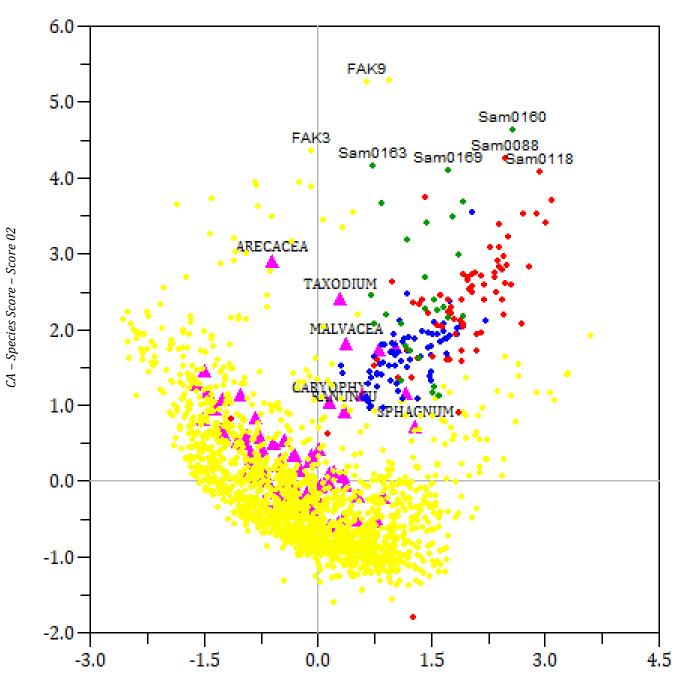


Figure 4 Principal component analysis for the modern and fossil data. Yellow dots = Florida, Blue dots = Noordwijk, Red dots = Petten and Green dots = Hank

Figure 5 represents the CA results for the modern and fossil data without *Pinus* and *Pinus haplox*. The red triangles indicate the optimum for the pollen taxa. *Arecaceae, Taxodium, Malvaceae, Caryophyllaceae, Ranuncul* and *Sphagnum* optima plot are distinct from the other taxa, whereby *Quercus, Betula, Cheno/Am* and *Picea* went to another corner. Samples further from this optimum point contain less of these species. The blue dots (Noordwijk) have the highest correspondence with the Florida dataset, green (Hank) seems to have the least resemblance to the modern dataset. The samples with the least resemblance are Sample 88, 118, 160, 163 and 169. The samples are outliers in this analyse because those samples are from the southern part of Florida and contain a lot of *Taxodium*. Those samples are relative warm in comparison to the other samples, maybe too warm for the Late Pliocene and Early Pleistocene for the Netherlands.

Correspondence graph modern and fossil data



CA – Species Score – Score 01 Figure 5 Correspondence graph for the modern and fossil data. Yellow dots = Florida, Blue dots =Noordwijk, Red dots = Petten and Green dots = Hank

4.2 Climate reconstruction

The performance of WA and WA-PLS are presented in Table 2 and Table 3 and the 'observed vs predicted' is presented in Appendix G and H, the performance of the MAT are presented in Table 4. Table 2 are the first results without removing any sample or taxa, Table 3 are without sample GPD1649, GPD1950, GPD1951 and GPD 1952 and the taxa *Pinus* and *Pinus haploxyon*. The samples have been chosen because of the outlying positions in the observation graphs (Appendix G). The inverse de-shrinking method is chosen for WA in stead of the classic method because the classic methods have a negative change in comparison to the inverse de-shrinking method. For the WA-PLS method both is chosen for component 3 because the %change to component 4 is minimal and to avoid overfitting in the model. Component 3 has a significant change to component 2 and a minimal difference with component 4. A small difference is observed between Table 1 and Table 2, only for the winter temperature reconstruction of WA is no rise in the bootstrapped R² value seen. All the others have a small rise in the value, so it seems that the reconstruction is a little bit better now.

	WA Total pre	cipitation		WAPLS Total precipitation			
Code	Boot R^2	%Change	RMSEP	Code	Boot R^2	%Change	RMSEP
WA_INV	0.512816		180.037	Comp. 1	0.51262		180.176
WA_Cla	0.513366	-34.9606	242.979	Comp. 2	0.5956	8.70763	164.487
WATOL_INV	0.402529		200.244	Comp. 3	0.642429	5.43333	155.55
WATOL_Cla	0.403677	-55.291	310.961	Comp. 4	0.668901	1.81805	152.722
				Comp. 5	0.661611	-0.0215982	152.755
	WA Winter te	mperature		WAPLS Winter temperature			
WA_INV	0.714195		3.83462	Comp. 1	0.713716		3.84124
WA_Cla	0.714653	-16.7989	4.47879	Comp. 2	0.807457	17.2186	3.173888
WATOL_INV	0.741713		3.70423	Comp. 3	0.829968	5.51694	2.94848
WATOL_Cla	0.742278	-14.6081	4.23534	Comp. 4	0.837909	1.7796	2.87854
				Comp. 5	0.838871	-0.0718837	2.8701

Table 3 Performance WA and WA-PLS after removing samples and Pinus and Pinus haploxyon

	WA Total pre	cipitation		WAPLS Total precipitation			
Code	Boot R ²	%Change	RMSEP	Code	Boot R ²	%Change	RMSEP
WA_INV	0.516889		179.384	Comp. 1	0.505986		179.558
WA_Cla	0.51772	-35.8083	243.618	Comp. 2	0.580981	7.71401	166.643
WATOL_INV	0.41889		195.82	Comp. 3	0.634105	5.97508	156.393
WATOL_Cla	0.419841	-51.5904	296.845	Comp. 4	0.652612	2.07448	152.999
				Comp. 5	0.655908	0.0725082	153.246
	WA Winter ter	nperature		WAPLS Winter temperature			
WA_INV	0.729023		3.73047	Comp. 1	0.728387		3.73866
WA_Cla	0.729342	-15.1814	4.29717	Comp. 2	0.807716	15.2154	3.16981
WATOL_INV	0.741262		3.371282	Comp. 3	0.8323	5.92074	2.98347
WATOL_Cla	0.741684	-14.3765	4.2466	Comp. 4	0.839519	1.63764	2.93669
				Comp. 5	0.841089	0.106758	2.93166

Percentile	Value	Value after removing samples
1	9.72514	9.71247
2	11.2313	11.2218
5	13.675	13.6659
10	16.0908	16.0823
20	19.4073	19.4015

Table 4 MAT percentile values for squared residual distance values before and after removing samples and Pinus / Pinus haploxyon

Table 4 shows the values for the MAT, those values indicate the goodness of fit. Fossil samples with a distance to the closest analogue larger than the 2nd and 5th percentile of the distances of the modern samples in the calibration dataset were classified as having 'no-close' and 'no-good' analogue, respectively (following e.g. Gouw-Bouman et al., 2019). The 20th percentile of MAT is presented in Figure 7 together with the reconstruction for the winter temperature and total precipitation. The age range for this graph is from 1.34 – 5.34 Ma years ago and include the Late Pliocene (cream colour) and the Early Pleistocene (blue/grey colour). The three different data sets are indicated by blue, red and green boxes, Noordwijk, Petten and Hank, respectively. The data points are marked by red circles and the errors bars are the small blue lines. The minimum and maximum temperature/precipitation is the same for the WA and WA-PLS to make a good comparison between both reconstruction methods. The percentage trees are the deciduous trees (Figure 8), the 100% sum is a calculation of deciduous trees, bisaccates, ferns, grasses, and other pollen types.

The resulting temperature range for the winter temperature is between -16 °C and 16 °C and the precipitation between 600 and 2100 mm per year. The minimum and maximum temperature lies between -12.47 °C and 13.77 °C for WA and -14.11 °C and 13.39 °C for WA-PLS. These optimum are reached at an age of 1.338 and 3.79 Ma years ago for WA and 1.338 and 3.217 Ma years ago for WA-PLS. The minimum and maximum precipitation lies between 924.4 and 1918 mm per year at an age of 1.34 and 3.79 Ma years ago for WA and 675.8 and 1529 mm per year, 2.372 and 1.597 Ma years ago for WA-PLS. During the Late Pliocene, the average temperature of this period is 4.12 °C/0.91 °C while the average temperature of the Early Pleistocene is below zero, -3.67 °C/-4.05 °C for the WA/WA-PLS reconstruction. A clear trend to a cooler world is observed at the transition to the Early Pleistocene for both reconstructions.

4.3 Optimum in the winter temperature and total precipitation

For the first 0.2 Ma year (2.38 – 2.58 Ma), the temperature is relatively low between -8.693 °C and 0 degrees for WA and -9.58 °C and 2.07 °C for WA-PLS. Also the percentage trees is low between 0-40%. However, the WA, WA-PLS and the percentage trees follow a similar pattern, when the percentage increases the temperature do the same. The next 0.2 Ma year (2.38 – 2.18 Ma) has a temperature range between -6.64 °C and 0.45 °C for WA and -10.31 °C and -0.31 °C for Wa-PLS. The percentage trees is between 2 and 87%, more difference than 0.2 Ma earlier.

4.3.1 Late Pliocene reconstruction

First, the general trend in the Late Pliocene is discussed, some differences between the WA and WA-PLS reconstruction are discussed. The highest temperature for WA is 13.77 °C, sample 162 (3.79 Ma) with a high concentration of *Pinus* and *Taxodium*. The highest temperature for WA-PLS is 13.39 °C, sample 160 (3.22 Ma), this sample has a peak abundance of *Osmunda* (86%), while a low concentration of *Taxodium* is found. Probably this is a sample from a sea-level low-stand (Dearing Crampton-Flood et al., 2020) and is an outlier. In general, the samples with

higher temperatures contain high concentrations of *Taxodium, Quercus* and *Polypodium*. The lowest temperature for WA is -1.26 °C, sample 157 (2.60 Ma), this sample contain a lot of bisaccates (*Abies, Pinus* and *Pinus haploxyon*) and some *Tsuga* and *Taxodium* (~5%). Sample 167 (4.75 Ma) is the coldest sample for the WA-PLS reconstruction with a temperature of -5.42 °C. This sample consists of some *Quercus* (3.2%), *Sphagnum* (2.4%) and *Taxodium* (85). The colder samples also contain a high abundance of *Abies,* while the warmer samples contain only *Pinus* and *Pinus haploxyon*. Other samples with relative high temperatures are 163, 169 and 172 (respectively 3.97; 4.97 and 5.43 Ma), these samples are 155, 161 and 171 (respectively 2.90; 3.24 and 5.17 Ma), these samples contain a higher concentration in *Ericaceae, Betula, Poaceae, Polypodium* and *Ulmus*.

The wettest sample for the WA reconstruction is 162 (3.79 Ma), also the warmest sample for the Late Pliocene reconstructed by WA. The same phenomenon is observed for WA-PLS, the same sample caused the wettest conditions (160, 3.22 Ma). The driest condition for WA is caused by sample 155 (2.94 Ma), this sample has a high diversity, including the bisaccates (41%), *Taxodium* (4%), *Ulmus* (3%) and *Polypodium* (21%). The driest sample for WA-PLS reconstruction is the same as the coldest conditions, sample 167. Other relative wet samples for both methods are 158, 163, 169 and 171 with ages 3.05, 3.97, 4.97 and 5.17 Ma years ago, respectively. These samples contain, except the bisaccates and the species mentioned above, *Alnus, Ericaceae, Osmunda* and *Poaceae*. Other relative dry samples are 156 and 165 (2.93 and 4.37 Ma years ago), with except for the bisaccates and the species mentioned above, only some *Ericaceae*.

4.3.2 Early Pleistocene reconstruction

The Early Pleistocene has more samples than the Late Pliocene, so more variation in the temperature and precipitation is observed. Only 7 samples are above zero degrees for the WA and 8 for the WA-PLS reconstruction. For WA, the highest temperature is observed for sample 97 (2.36 Ma) with 5.39 °C. The sample contains a high amount of Polypodium (50%) and Betula Osmunda and Cheno/Am, all 13%. The highest temperature for WA-PLS method is also observed in sample 97, 5.29 °C. All the other samples with a temperature above zero, for both WA and WA-PLS, contain a high amount of *Polypodium* and do not contain as many bisaccates as during the Late Pliocene. Other pollen species with relative high concentrations in the warmer samples are Alnus, Poaceae, Ericaceae, Sphagnum, Cyperaceae and Tsuga. The lowest temperature for WA method is caused by sample 74 (2.29 Ma), this sample contain 50% Betula and 50% Pinus. Another sample with a relative low temperature is sample 63 (-10.03 °C; 2.23 Ma), this sample also contain a high amount of Pinus (36%) and Ericaceae (22%). For WA-PLS, again sample 74 caused the lowest temperature (-14.11 °C), other cold samples are number 85 and 142 (2.32 and 2.57 Ma). Number 85 contain 50% Polypodium and 30% Ericaceae, number 142 also contains a lot of Pinus (29%), Polypodium (18%) and some Picea (22%). Other pollen species which are associated to colder conditions are Alnus, Cyperaceae, Poaceae and Quercus.

The precipitation range is almost the same for the Early Pleistocene as during the Late Pliocene. The wettest sample for the WA method is number 97 (2.36 Ma) with 1433 mm per year. This sample is again the same sample as the warmest sample for the WA method, the same phenomenon as during the Late Pliocene. For WA-PLS, sample 88 is the wettest sample with 1529 mm per year at an age of 2.34 Ma years ago. This sample is not the same as the wettest sample for the WA-PLS method. However, this sample also contains a high amount of *Polypodium* (86%), other pollen species in the sample are *Ericaceae* (5%), *Poaceae* (2%) and *Alnus* (2%). Other wet sample contain lot of *Ericaceae*, *Pinus* and *Sphagnum*. The driest sample for the WA-method is sample 89 (2.34 Ma), this sample is very close to the wettest sample of the

WA-method. The most important pollen species for this sample is *Poaceae* (92%). The WA approach produces in general warmer and wetter reconstruction than WA-PLS. The WA-PLS method reconstructs an overall drier environment than the WA-method, with values around the 700 mm per year in comparison to WA with ~950 mm per year. The driest sample for WA-PLS is 32 (1.84074 Ma), this sample contain *Pinus* (26%), *Poaceae* (18%) and *Artemisia* (15%). Other pollen species which cause dry conditions are *Betula, Cheno/Am, Cyperaceae* and *Quercus*.

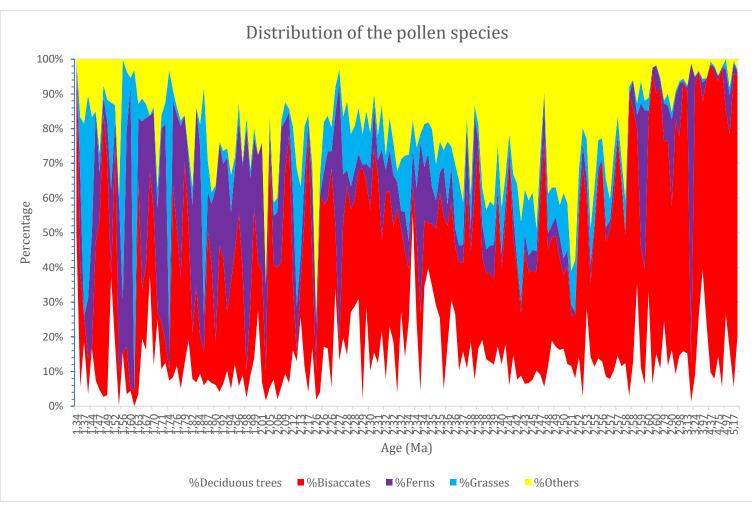
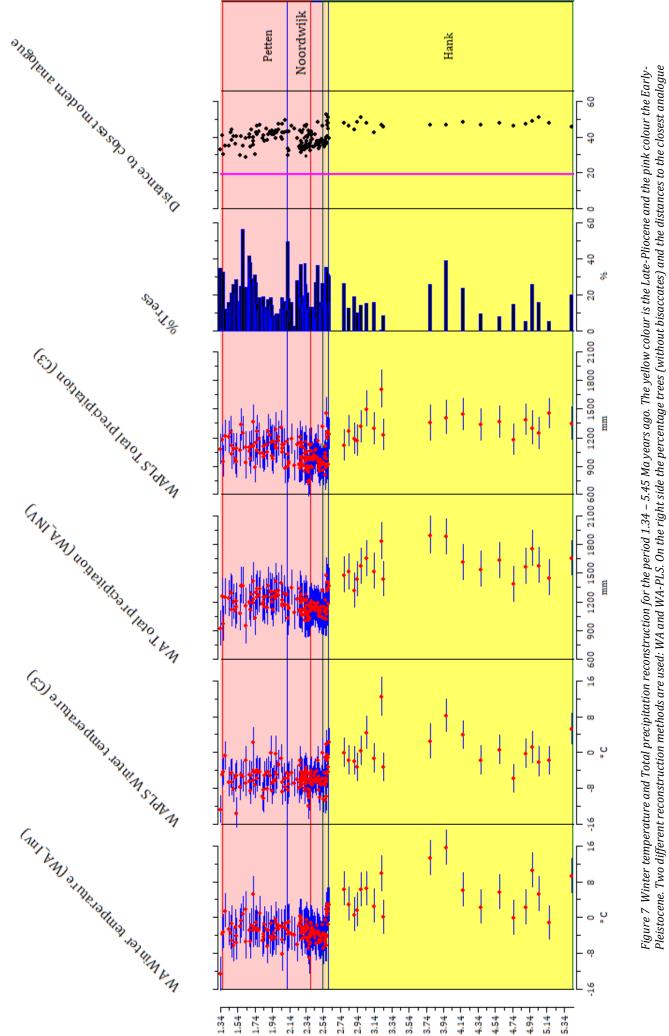


Figure 6 Distribution of the pollen species with Age (Ma). Deciduous trees are Acer, Alnus, Anacardia, Arecaceae, Betula. Carpinus, Carya, Celtis, Corylus, Fagus, Fraxinus, Juglans, Liquidambar, Nyssa, Quercus, Salix, Taxodium, Salix and Ulmus. The bisaccates are Abies, Picea, Pinus, Pinus haploxyon and Tsuga. The ferns are Osmunda and Polypodium. The grasses are Poaceae.



5. Discussion

5.1 How accurate are the climate reconstructions?

It is important to know if the reference dataset is a valid dataset to use for this kind of climate reconstructions and period for the Netherlands. Two options are used to check whether the reconstruction made is correct. The first option is to compare the reconstruction with the percentage of deciduous trees, the section option is to check our reconstructions with known data from other reconstructions from the same period.

5.1.1 Comparison with percentage deciduous trees

From Marshak, 2010, pp 795, "Deciduous forest are common in warmer climates and coniferous forests are more common in colder environments". Bisaccates are more dominant during colder climates and grass pollen and deciduous forests are more dominant during warmer climate periods (e.g. Marshak, 2010, pp 795 – 795). However, during the Pliocene bisaccates are more common in warmer areas due to well developed soil. The precipitation graph should follow roughly the same curve as the temperature graph because during warmer periods more precipitation is expected. Warm air contain more water vapor, the warm air rises and cools. As a result, the water vapor condenses and the small droplets form large clouds which cause the precipitation.

The percentage deciduous trees (Appendix M) is showed in Figure 7, these trees are: *Acer, Alnus, Anacardia, Arecaceae, Betula, Carpinus, Carya, Celtis, Corylus, Fagus, Fraxinus, Juglans, Liquidambar, Nyssa, Quercus, Salix, Taxodium, Tilia* and *Ulmus*. Figure 6 shows the distribution of the pollen taxa with age (Ma). The Pliocene had a higher percentage of deciduous trees, bisaccates, ferns and grasses than the Pleistocene, a new indication of a warmer period during the Pliocene. The figure shows the trend into a colder period whereby the 'others' increase. The WA reconstruction (winter temperature) for the Late Pliocene follows the pattern of the extreme in the %trees graph. The precipitation graph follows roughly the same pattern as the temperature graph. The WA method fits better in the %trees graph than the WA-PLS method for both reconstructions. The peak, observed around 3.217 Ma, is not observed in the %trees graph. This peak is caused by a high amount of *Osmunda*. *Osmunda* is an indication of wet and warm conditions during the Late Pliocene. *Osmunda* is a terrestrial plant mainly found in temperate or sub-tropical regions, this indicates a relative more wat and warmer conditions during this period with less deciduous trees.

The Pleistocene has more samples than the Pliocene, because of the high number of samples, the Pleistocene is subdivided into different parts to observe the data closer. In general, the WA-PLS reconstruction follows roughly the same pattern as the %trees graph. Two peaks are observed closer because the peaks are relatively high in AP% in comparison with the reconstruction temperature. The peaks are caused by sample 23 and 26, these two sample contain *Alnus* (11/16%), *Betula* (9/8%), *Poaceae* (8/9%), *Polypodium* (9/165) and *Quercus* (25/-5).

The period (2.18 – 1.98 Ma) has a temperature range between -7.929 °C and -0.755 °C for WA and -6.813 °C and 3.056 °C for WA-PLS. The tree percentage varies between 5 and 87% with two peaks which are not observed in the WA and WA-PLS reconstruction. Also in the precipitation graphs, these increases in %trees are not observed. The two peaks are caused by sample 3 and 131 and contain mainly ferns, like *Sphagnum* (18/5 %) and *Polypodium* (5/13 5) and grasses like *Poaceae* (20/5 %). The deciduous trees found in these samples are *Alnus, Betula* and *Corylus*. For the period 1.98 – 1.78 Ma, one outlier is observed caused by sample 103. This sample contain mainly ferns like *Polypodium* and *Sphagnum*, also some *Betula* is observed. The period, 1.78 – 1.58 Ma, follows roughly the same pattern as the reconstructed data. One data point has a relative high percentage of deciduous trees and a very low temperature of -8 °C. This datapoint

is sample 93 and has a high percentage of ferns (*Osmunda, Polypodium*) and some deciduous trees (*Alnus, Betula, Quercus*). The youngest part of the Pleistocene has one peak with high percentages of deciduous trees and a relative low temperature/percentage. This is caused by sample 83 with *Alnus* (25%), *Quercus* (12,5%), *Ericaceae* (12,5%), *Poaceae* (25%) and *Polypodium* (25%). In general, the WA-PLS temperature range is higher than the WA reconstruction, however, the same mismatches with the %deciduous trees are observed. The most common pollen taxa are *Polypodium*, *Poaceae* and *Sphagnum* which are found in the outlying samples. The combination of these pollen taxa causes probably a higher temperature than the %deciduous trees predicts.

5.1.2 Comparison with other datasets

The reconstructed data from Hank is compared with a temperature reconstruction from Dearing Crampton-Flood et al., (2020). Dearing Crampton Flood et al., (2020) reconstructed the mean temperature by using three independent organic temperature proxies for sea surface and terrestrial temperatures based on different lipid biomarkers. A multi-proxy record of Pliocene climate change is made based on the sedimentary record for the borehole of Hank. This borehole is the same as used in this guided research.

The reconstructed data from this guided research follows roughly the same pattern as the reconstruction from Dearing Crampton-Flood et al., (2020), however, some differences are observed (Figure 9). The differences are observed between 2.58 – 2.93 Ma and 3.32 – 4.37 Ma. An important note to this comparison is the amount of datapoints, the reconstruction from

Dearing Crampton-Flood et al., (2020) contain more datapoints than this research had used. More datapoints makes the reconstruction more precise and accurate. Some increase of decrease in temperature are not registered in our reconstructions because the datapoints are not present. Because of the lack of datapoints, the differences described below can be explained. For the period 2.58 – 2.98 Ma, the reconstructed GDGT data from Dearing Crampton-Flood et al., (2020) varies between 4.4 – 7.9 °C and the reconstructed data in this guided research fluctuates between 11.07 – 16.84 °C. During the other period (3.32 - 4.37 Ma), the reconstruction of Dearing Crampton-Flood et al., (2020) fluctuates between 3.91 - 12.4 °C and a variety of 10.15 – 24.95°C for this reconstruction. It is hard to conclude from this comparison which transfer function is better to use because both, WA and WA-PLS, following the same pattern. WA seems to fit a little better than WA-PLS.

Another dataset which used to compare the reconstructed data is the data from Zagwijn (1992). Zagwijn (1992) made the best guess for the Late Pliocene and Early Pleistocene for the Netherlands, however, his data is not validated, and the age model is old. Zagwijn (1992) used the minimal summer temperature following the method of Iversen (1944) (Figure 8). His guess for the mean temperature in July varies from 0 to 20 °C and during the Late Pliocene the

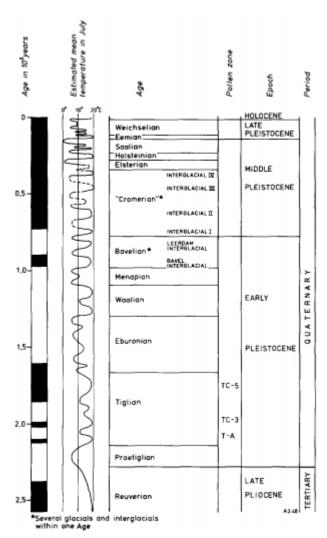


Figure 8 Temperature reconstruction from Zagwijn (1992)

summer temperature does not drop below the 10 °C. Zagwijn' temperature guess for the Early Pleistocene is warmer during the transition to the Early Early Pleistocene and the same during the Early Pleistocene. The data from Zagwijn (1992) indicates that our temperature reconstruction is in the right temperature range.

Comparing our reconstructed data with other datasets indicate that the temperatures are roughly in the same temperature ranges. For further indication it is good to compare the data with more datasets and to use more data points for a better comparison.

5.1.3 The differences between WA and WA-PLS

WA-PLS is a modification of WA, however, despite the modification WA is the most used method for paleo-environmental reconstructions. WA-PLS wants to fit all the variation into the reconstruction, while WA just showed the reconstruction without fitting it all. The WA-PLS method indicates a higher temperature for both Late Pleistocene and Early Pleistocene. The WA indicates more precipitation for the Late Pliocene and less precipitation for the Early Pleistocene in comparison with the WA-PLS method. Two methods are used to indicate which method is better to use for this timespan and region. The first method is the percentage deciduous trees in the samples. The WA seems to fit better in the older data from the Late Pliocene, while the younger data better fits in the WA-PLS reconstruction.

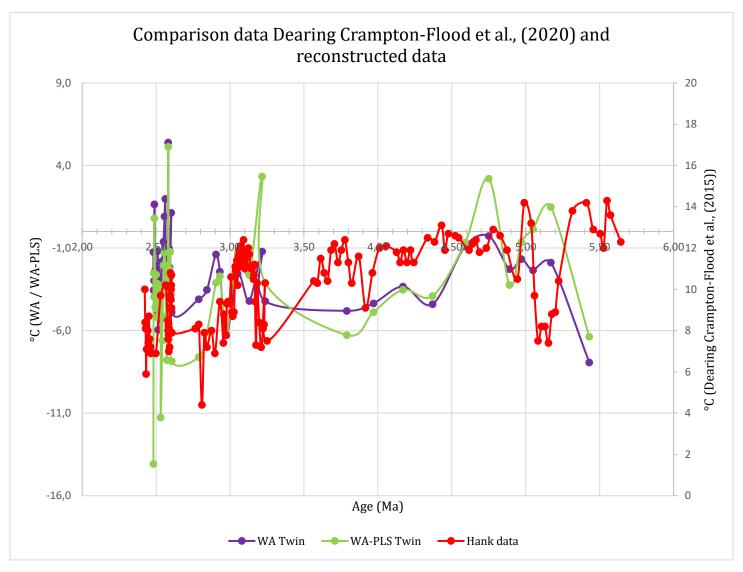


Figure 9 Comparison reconstructed WA and WA-PLS method with data from Dearing Crampton-Flood et al., (2020). The winter temperature is purple (WA) and green (WA-PLS). The data from Dearing Crampton-Flood et al., (2020) is red.

An observation was the existence of the ferns during the periods of high percentage of deciduous trees with colder temperatures. The ferns seems to create a colder temperature with an average precipitation pattern. Especially *Polypodium*, a spore, has a high occurrence during the colder temperature periods with a high occurrence of deciduous trees. This observation was also done in Chapter 4 Results, whereby the colder temperatures were characterised by the occurrence of *Polypodium* for both Late Pliocene and Early Pleistocene. A suggestion should be to do a run without the ferns for future research to indicate whether the ferns actually bias the temperature reconstructions.

The second method is the comparison with known data from Dearing Crampton-Flood et al., (2020). It is hard to indicate which method is better because of the lack of datapoints for our reconstruction. For the older data, the WA seems more likely because the extremes are more visible in Figure 9 than by the WA-PLS method. Mainly during the period 3.50 – 4.50 Ma, a low resolution have been reconstructed for this guided research and it is hard to make any conclusions.

5.2 How accurate is the data from Whitmore et al., (2015)? Sample GPD1649, GPD1950, GPD1941 and GPD1952 were removed from the dataset because of the outlying position in the PCA and CA graph. GPD1649 contain a high percentage of *Pinus, Quercus* and *Arecaceae*. GPD1950 contain a high percentage of *Abies, Betula, Pinus, Polypodium* and *Quercus* and sample GPD1952 *Betula* and *Quercus*. The removed samples have a latitude between 29° and 35°. *Quercus* is a dominant pollen species in between these latitudes. Also sample FAK3 and FAK 9 are removed, these two samples contain a high amount of *Taxodium* and *Polypodium*. These two samples are relative warm, known by the high percentage of *Taxodium*. This could be a reason why these two samples were outliers in the PCA and CA. As indicated by Table 2 and Table 3, not much changed after removing these six samples. Removing these samples makes the reconstruction a little bit precise but do not change enough to see large changes.

A specific part of the dataset of Whitmore et al., (2015) was used for this reconstruction. The limit of 51° N is made because of the temperature range from the pollen species. Going further north than 51° N, the climate is more similar to montane and continental Europa instead of the lowland sea climate in the Netherlands during the Late Pliocene and Early Pleistocene. The real problem is the occurrence of similar types like *Juniperus* in cool areas that (within the Cupressaceae type) had a broader/warm distribution in the Pliocene in Europe. Also some cold pollen species do not occur in the Florida region, like *Abies* and *Picea*, now relatively a warm and cold pollen species. In the dataset of Hank they occur at the same time, an indication that some plants changed their climatic characteristics which makes it hard to compare the fossil datasets with the modern ones. Because these two species do not overlap in the modern dataset (Appendix A). Other pollen taxa which have more correspondence in the fossil dataset instead of the modern one were Arecaceae, Caryophyllaceae, Malvaceae, *Ranunculus, Sphagnum* and *Taxodium*, a non-analogue condition. These are the taxa that characterize the right/upper side of the CA plot.

The occurrence of *Betula* and *Quercus* at the same time occurs only in the upper part (Early Pleistocene), during the Late Pliocene they barely occur together. One of the options is the occurrence of different *Quercus* species. Another observation in the pollen species was the occurrence of the bisaccates. After analysing the PCA and the CA, *Pinus*, and *Pinus haploxyon*-type are removed from the dataset because they were outliers in the graphs. It was suspected that this pollen taxon was located nearby the other bisaccates. Another reason to remove *Pinus*

was because of the dominant occurrence in the samples and the strong bi-modal distribution, this could give a disturbance in the reconstruction has too much emphasis on the reconstruction.

During the Late Pliocene and Early Pleistocene, Pinus occurs in a different environment than present. In the Florida data, *Pinus* only has a decrease at a latitude of \sim 38 °. During the Early-Pleistocene, Pinus, Picea and Taxodiaceae pollen grains are dominant with Fagus, Quercus, Zelkova and Ulmus pollen grains (Fujiki and Ozawa, 2008). In the Early Pleistocene, pollen grains (Liquidambar, Tsuga, Pterocarya and Carya) disappeared (e.g. Svenning, 2003). Pinus, Picea and Taxodiaceae pollen grains were still dominant and Tsuga pollen increased (Fujiki and Ozawa, 2008; Svenning, 2003). The climate became colder from the Late Pliocene to the Early Pleistocene and *Pinus, Tsuga* and *Abies* grew in the highland areas at altitudes from 1500 to 3000 m above sea level (Fujiki and Ozawa, 2008). In Florida, the Pinus occur in warm lowlands, while during the Late Pliocene and Early Pleistocene *Pinus* occur in the colder highlands in Europe. *Pinus haploxyon* occurs in the warmer coastal lowlands during the Pliocene. Another aspect was the differences between the occurrence of Picea, Pinus and Abies. During the Pliocene Picea and Abies had a high occurrence in overall, currently *Picea* and *Abies* occurring in the higher colder areas. Pinus has a higher abundance during warmer temperatures and wetter conditions while *Abies* abundance results in colder temperatures due to its present distribution in that habitat. In general, the bisaccates had different environments during the Pliocene and the Pleistocene for Europe and America. This makes it hard to make a good reconstruction with these taxa in the reconstruction, for future research it seems good to indicate which kind of the bisaccates occur in the different areas and to specify their climate characteristics.

Another striking aspect is the occurrence of *Osmunda*, *Osmunda* is an indicator of warm and wet conditions during the Late Pliocene. The Whitmore data contain two samples with a high amount of *Osmunda*, these samples are GPD165 and GPD591. The average temperature for this sample is 16.23 °C with a total precipitation of 1105 mm/year. The average winter temperature is 6.7 °C degrees, in comparison to other samples with less *Osmunda*, this temperature is low. GPD591 has an average temperature of 3.34 °C degrees and a winter temperature of -3.8 °C degrees.

For future research, the dataset of Whitmore et al., (2015) is a good dataset to use, however, it is important to know the climatic characteristics from the dominant pollen species for now and during the Late Pliocene and Early Pleistocene.

6. Conclusions

The aim of this guided research was to reconstruct the climate during the Late Pliocene and Early Pleistocene for the Netherlands based on pollen assemblages data with quantitative paleoclimatic reconstruction methods. The reconstruction for this timespan is based on three different terrestrial datasets from the Netherlands. The datasets are in Noordwijk, Petten and Hank. Two different pollen-climate transfer functions were calculated using two way Weighted Averaging (WA) with inverse deshrinking and 3- component Weighted Averaging Partial Least Squares (WA-PLS) regression. Calibration set species data values were square-root transformed for WA and WA-PLS regression to reduce noise in the data.

Two different climate parameters are reconstructed: winter temperature and total precipitation. *Pinus* and *Pinus haploxyon* have a very high presence in all the data samples, so these pollen species are removed from the reconstruction, substantiate by the PCA and CA. *Pinus* and *Pinus haploxyon* were both outliers in these graphs.

Two methods were used to indicate which technique is better to use for climate reconstruction for this time span. The first method was to compare the reconstruction with the percentage deciduous trees and the second one was to compare the reconstruction with known data. Concluded from the first method, the WA method is better to use with older data (Late Pliocene), WA-PLS can be used for younger data. WA-PLS wants to fit all the variation in the reconstruction, while this is for older data not the right way to reconstruct. As we see in our data, some pollen species do not exist anymore or have another climate characteristic in comparison with their present pollen species. If WA-PLS tries to fit all the variation in the reconstruction, some mistakes are going to be made in the reconstruction.

The other method was to compare our data with known data from Dearing Crampton-Flood et al., (2020) and to indicate if the same trends from Zagwijn (1992) are observed for the Pleistocene. Again, WA seems to fit better in the reconstruction, however, it is hard to make a good conclusion because of the differences in datapoints. For a long time, Zagwijn (1992) has made the best estimate for the temperature range in the Netherlands for the Late Pliocene and Early Pleistocene. Zagwijn (1992) made a summer temperature reconstruction, while our data is a winter temperature reconstruction. The data of Zagwijn (1992) is compared with the average temperature reconstruction. The Late Pliocene data seems right, the temperature is in the same range as Zagwijn (1992) reconstructed. The transition to the Early Pleistocene seems too cold, however, the reconstruction from Zagwijn (1992) is older, however, not wrong, and is not dated accurately.

For further research, it would be better to use more fossil datapoints for a better comparison with other data. It is hard to indicate if the data is right because, especially for the Late Pliocene, not enough data points are available. Also, another dataset could be used as calibration set, for example for the South-Eastern part of Asia. It is also good to know the difference between the climate characteristics of some pollen species now and during the Late Pliocene and Early Pleistocene. Also, the influence of the Ferns on the climate was a striking element found in this research. Another method which could be used for the reconstruction is CREST, a method from Chevalier. The CREST method is related to a Bayesian approach combines presence-only occurrence data and modern climatology's to estimate the conditional response of a given taxon to a variable of interest.

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List with figures

Figures:

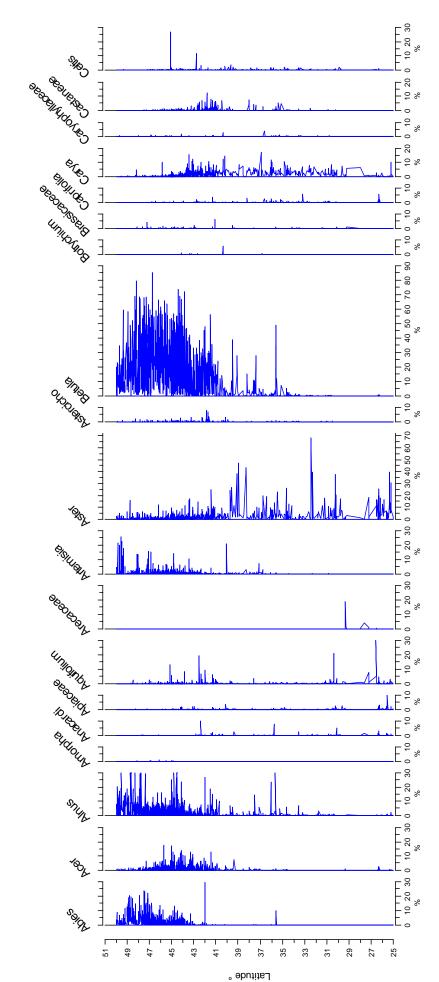
- 1. Conceptual diagram describing MAT. Taxon 1 prefers humid climates (green) and taxon 2 prefers drier conditions (brown). Three fossil assemblages (diamonds) are compared to a modern data set, each on associated with a humidity value (coloured circles). The five closest analogues (white circles) are retained and averaged to estimate past conditions (Chevalier et al., 2020)
- 2. Conceptual diagram describing WA-PLS. The black part indicates all taxa which are not considered in the example. On the right, the ecological gradient is showed, the modern assemblages. Climate values are extracted from each modern sample to derive the climate optimum of the taxa, the small circles on the right (Chevalier et al., 2020)
- 3. Location of the fossil datasets, from north to south: Petten, Noordwijk and Hank
- 4. Principal Component Analyse for modern and fossil dataset. Yellow dots = Florida, Blue dots = Noordwijk, Red dots = Petten and Green dots = Hank.
- 5. Correspondence graph for the modern and fossil dataset. Yellow dots = Florida, Blue dots = Noordwijk, Red dots = Petten and Green dots = Hank
- 6. Distribution of the pollen species with Age (Ma). Deciduous trees are Acer, Alnus, Anacardia, Arecaceae, Betula. Carpinus, Carya, Celtis, Corylus, Fagus, Fraxinus, Juglans, Liquidambar, Nyssa, Quercus, Salix, Taxodium, Salix and Ulmus. The bisaccates are Abies, Picea, Pinus, Pinus haploxyon and Tsuga. The ferns are Osmunda and Polypodium. The grasses are Poaceae.
- 7. Winter temperature and Total precipitation reconstruction for the period 1.3375 5.3375 Ma years ago. The creme colour is the Late Pliocene and the blue/grey part is the Early Pleistocene. Two different reconstruction methods are used: WA and WA-PLS. On the right side the percentage trees (without bisaccates) and the distance to the closest analogue are showed. The red dots are the data points, and the blue lines are the errors bars. The blue box indicate the samples of Noordwijk, the red box of Petten and the green box of Hank.
- 8. Temperature reconstruction from Zagwijn (1992)
- 9. Comparison reconstructed WA and WA-PLS method with data from Dearing Crampton-Flood et al., (2020). The winter temperature is purple (WA) and green (WA-PLS), the average temperature is yellow (WA) and blue (WA-PLS). The data from Dearing Crampton-Flood et al., (2020) is red.

Tables:

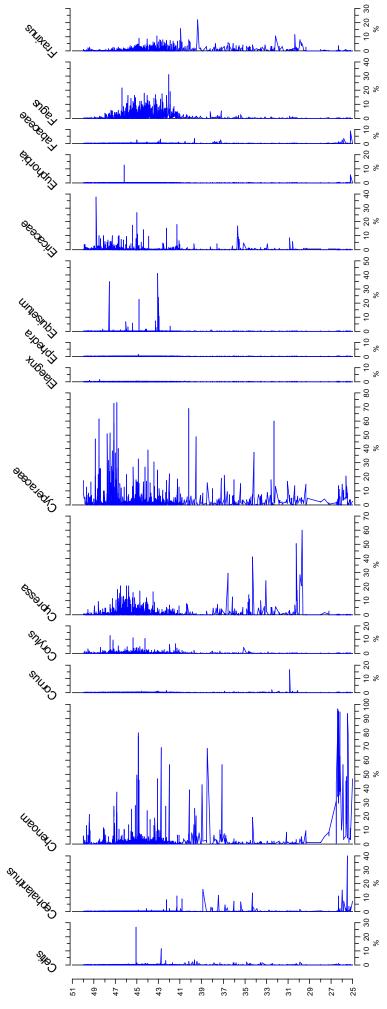
- 1. Performance WA and WA-PLS
- 2. Performance WA and WA-PLS after removing samples and Pinus and Pinus Haploxyon
- 3. MAT values before and after removing samples and Pinus and Pinus Haploxyon

Appendices

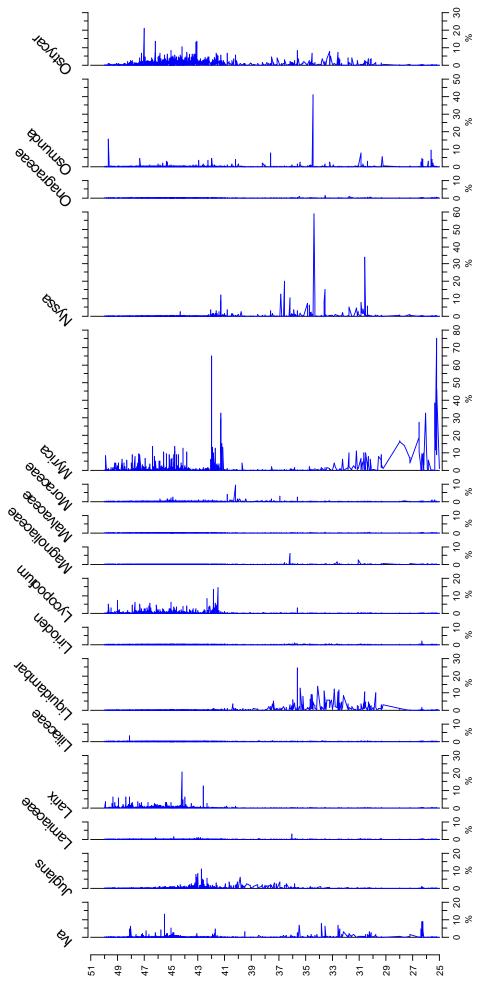
- A. Pollen diagram modern dataset Florida
- B. Pollen diagram fossil dataset Noordwijk
- C. Pollen diagram fossil dataset Petten
- D. Pollen diagram fossil dataset Hank
- E. Dominant pollen species for fossil datasets with age
- F. Principal Component Analyse Modern and Fossil data
- G. Correspondence graph with all the data
- H. Observation graphs WA and WA-PLS
- I. Observation graphs WA and WA-PLS with removed samples
- J. First climate reconstruction
- K. Calibration sample number age
- L. Fossil pollen species combined with modern pollen species
- M. Percentage trees & bisaccates

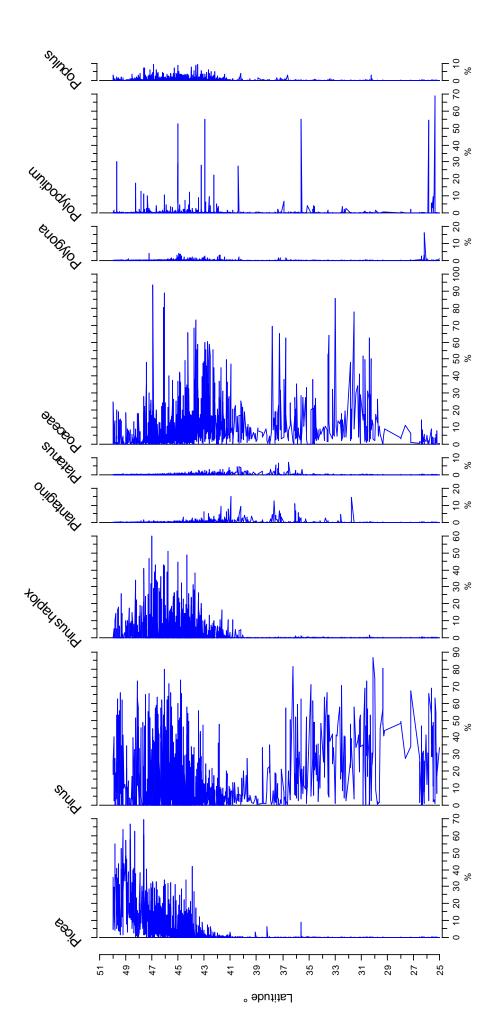


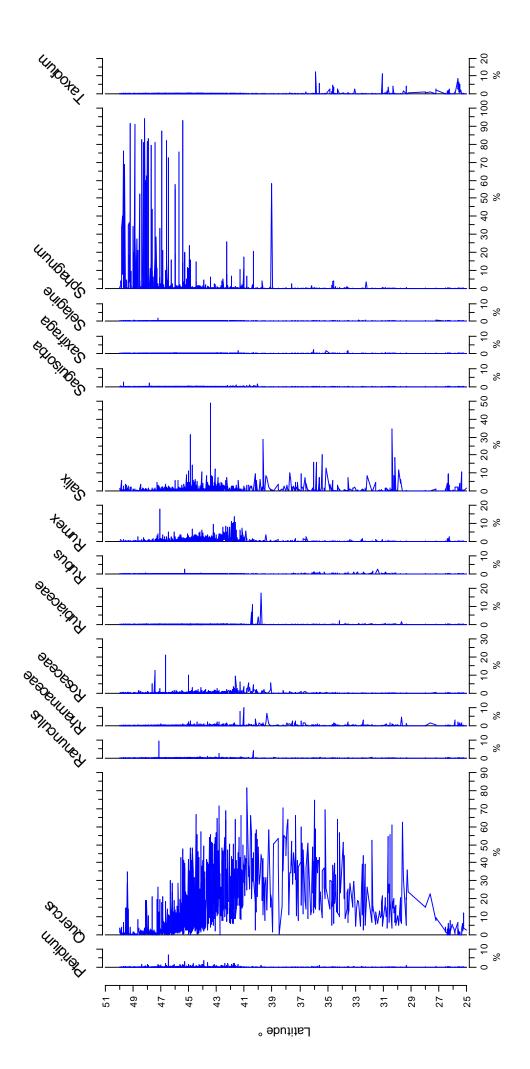
A. Pollen diagram modern data set Florida

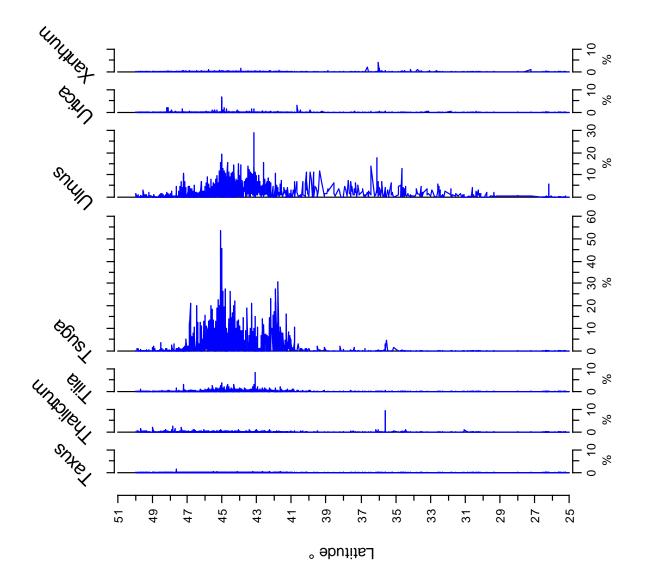


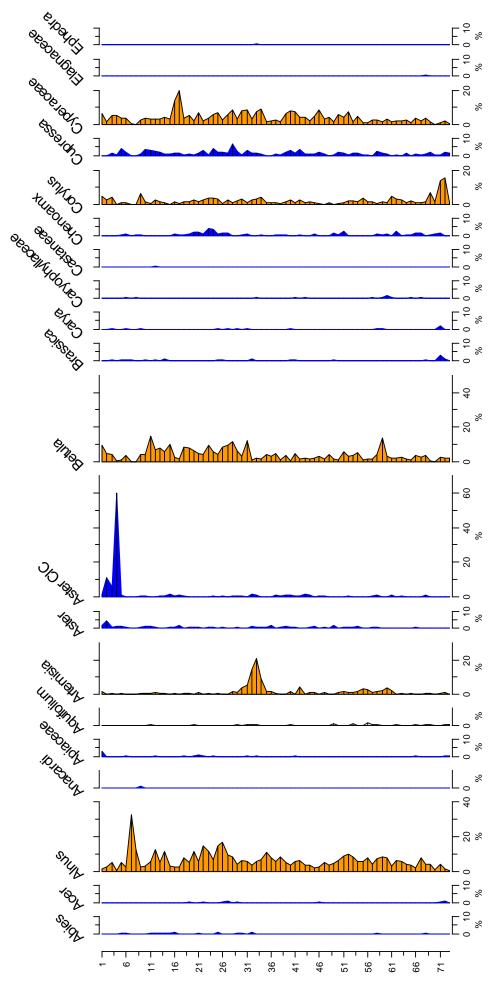
° sbutite C



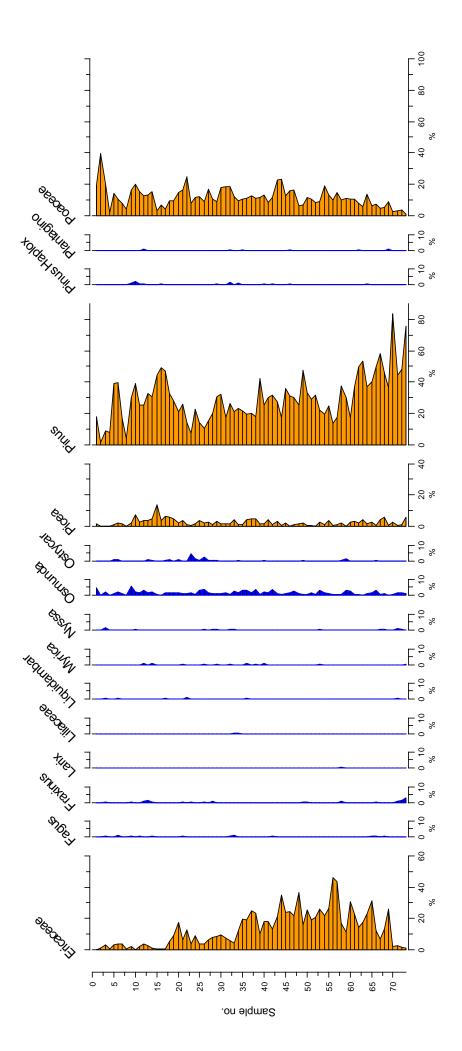


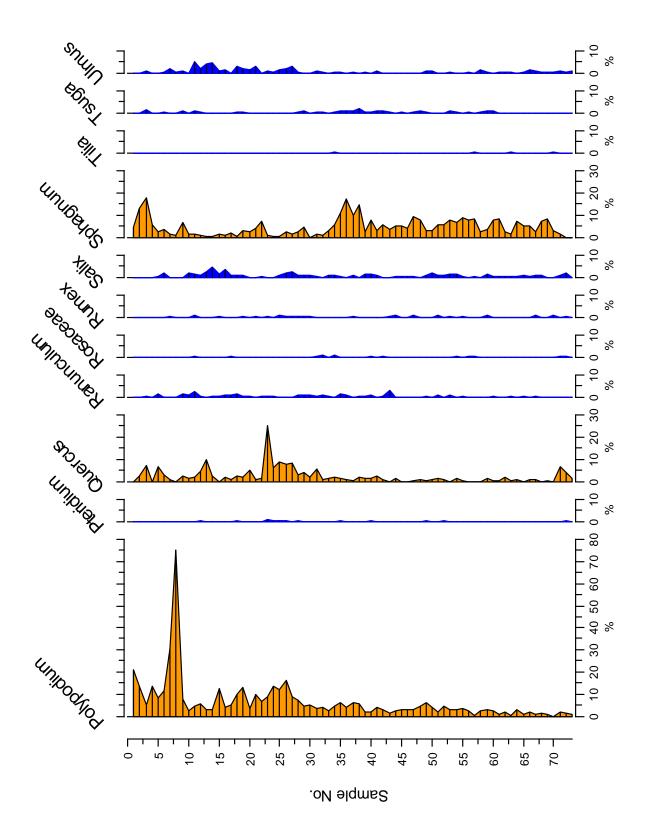




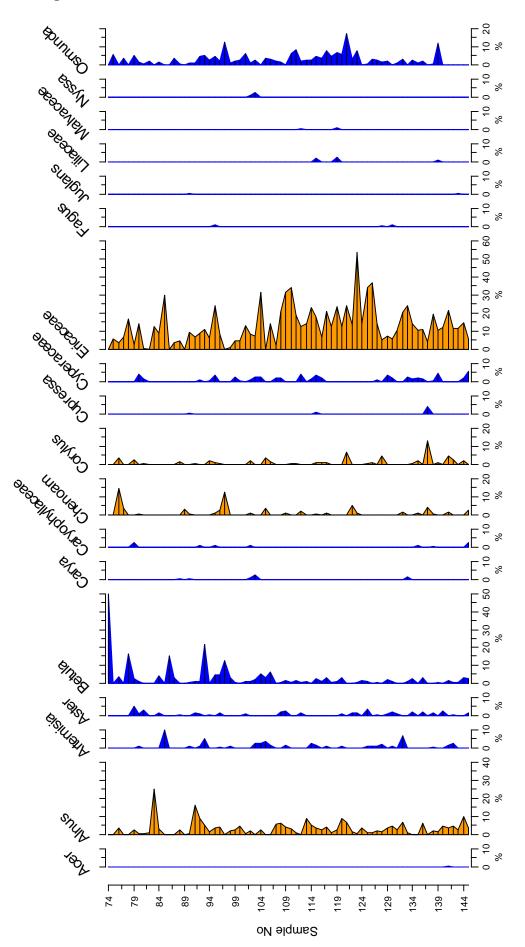


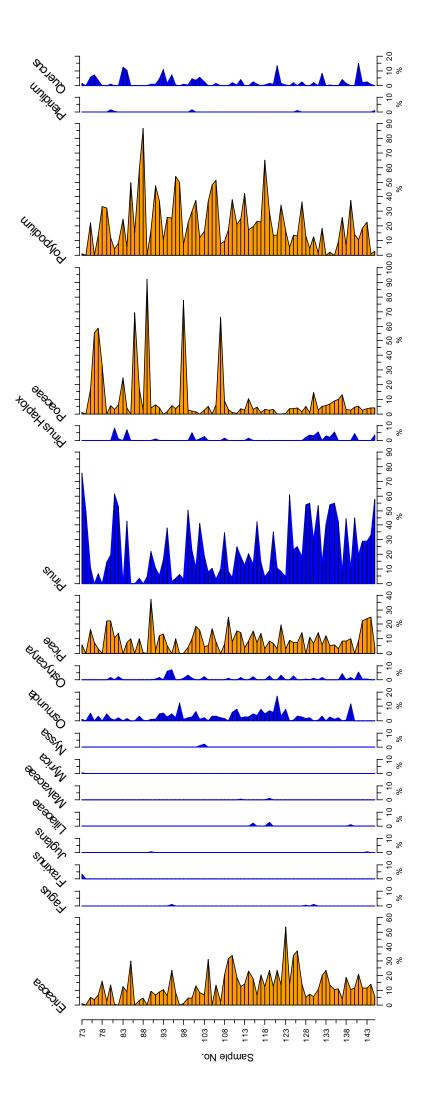
B. Pollen diagram fossil dataset Noordwijk

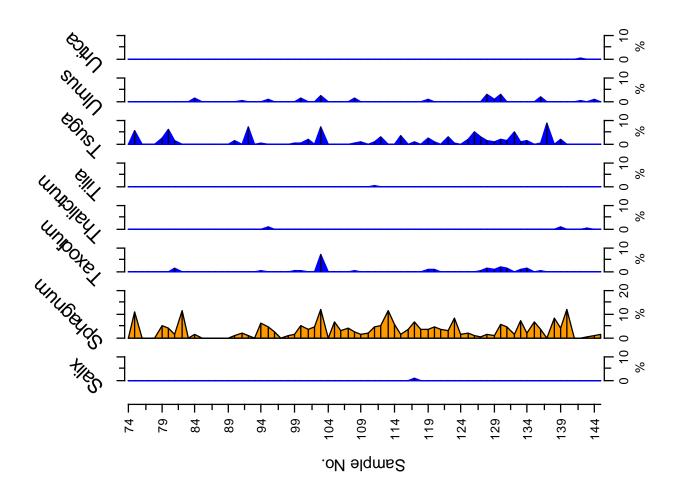




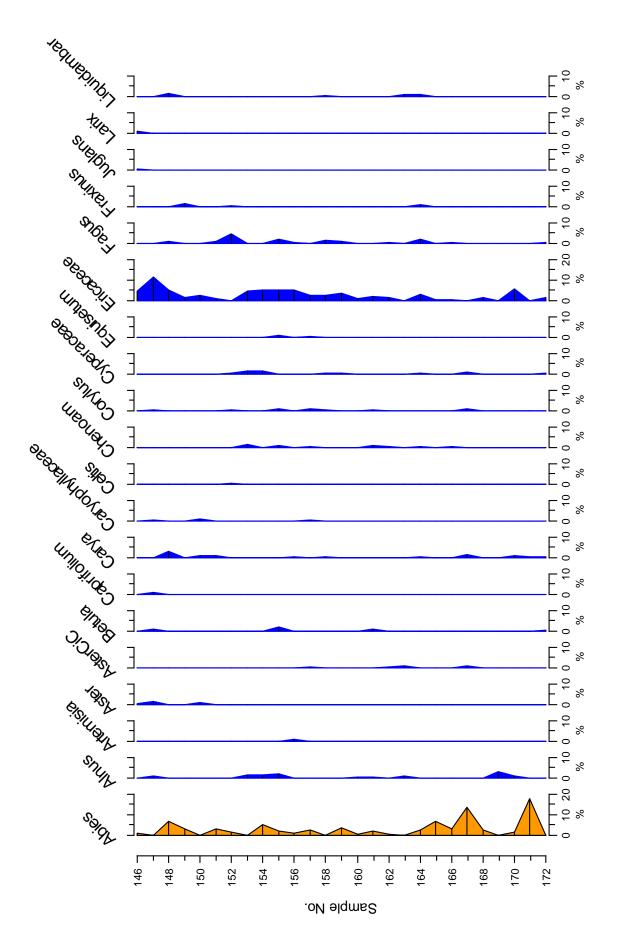
C. Pollen diagram Petten

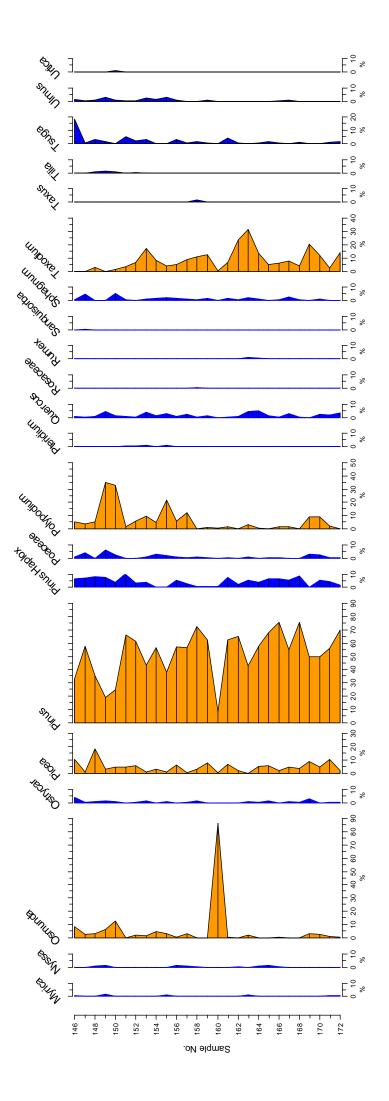


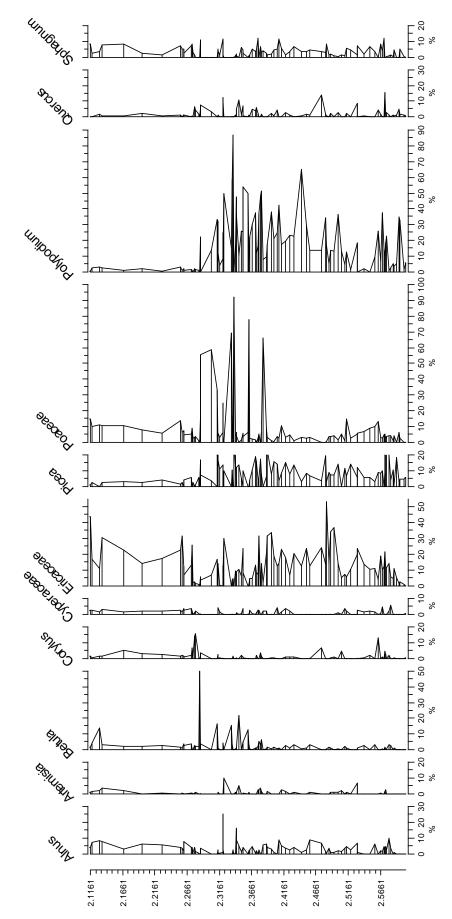




D. Pollen diagram Hank

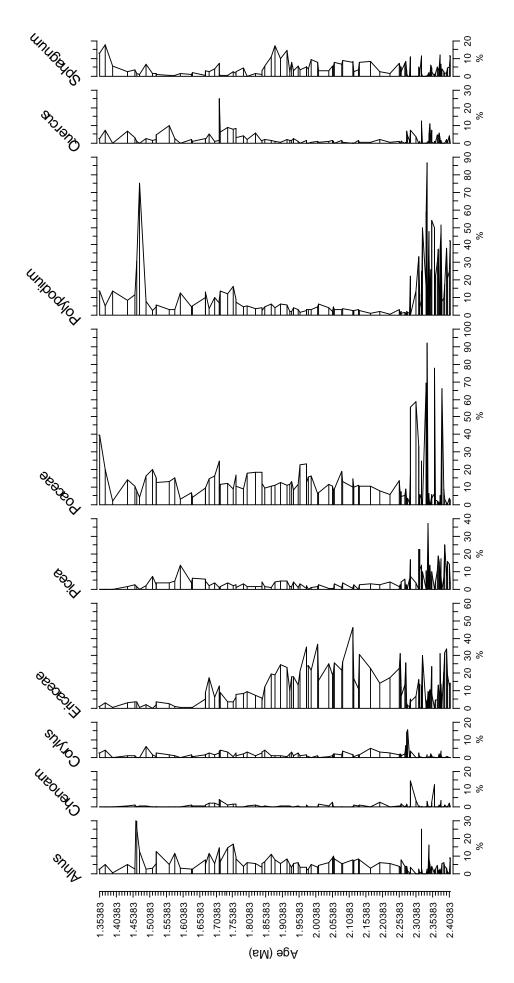




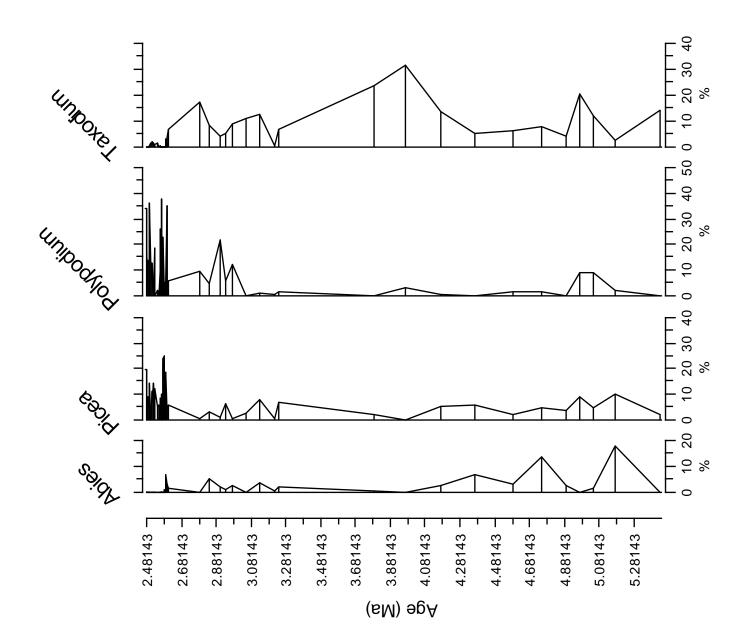


E. Dominant pollen species for fossil data sets with age Noordwijk

Petten

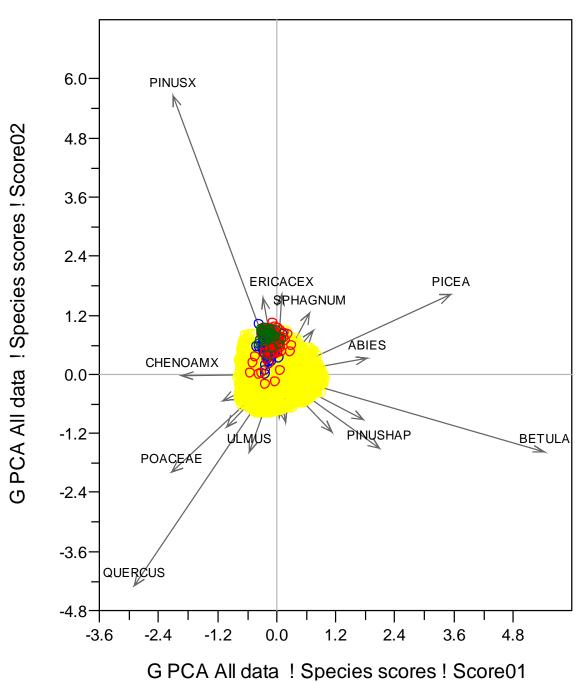






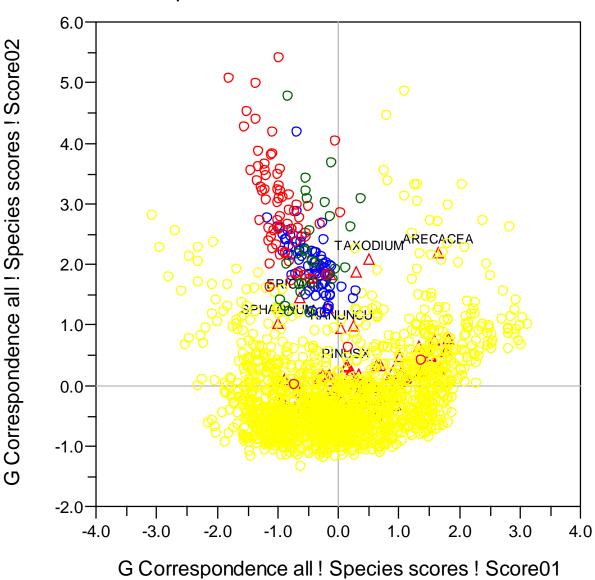
F. Principal Component Analyse Modern and Fossil data

Principal Component Analyse including all the data

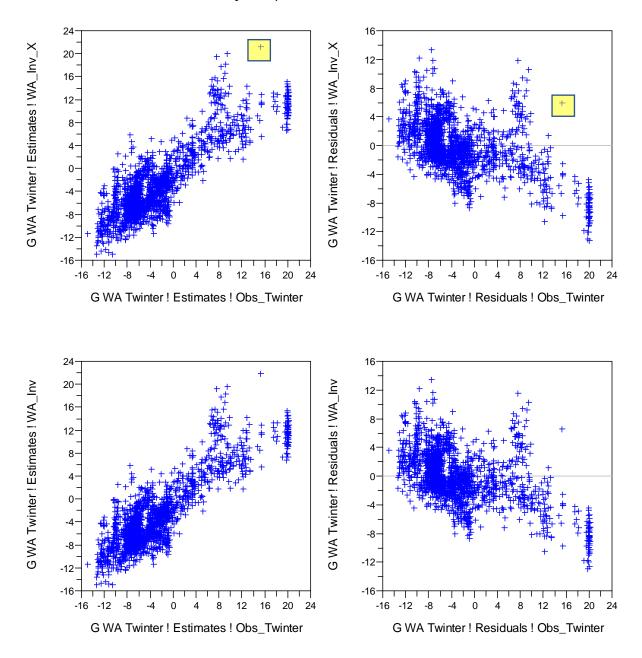


PCA All data Modern + Fossil

G. Correspondence graph with all the data



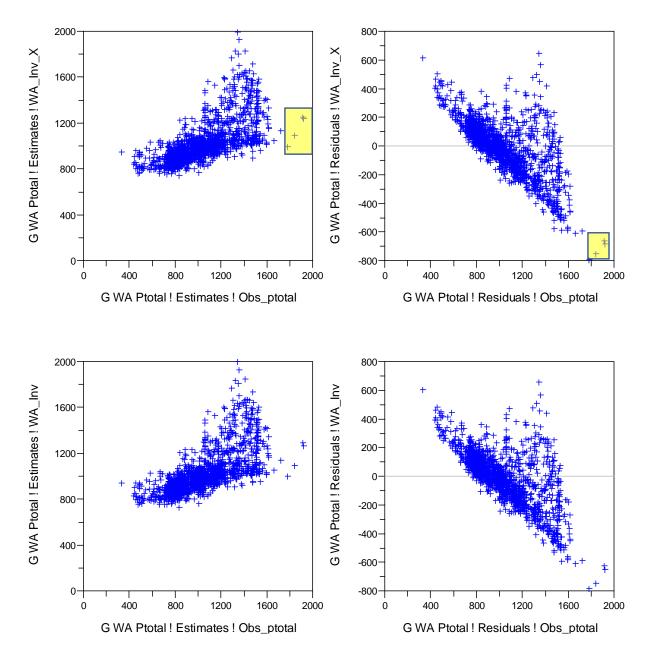
Correspondence Modern + Fossil data



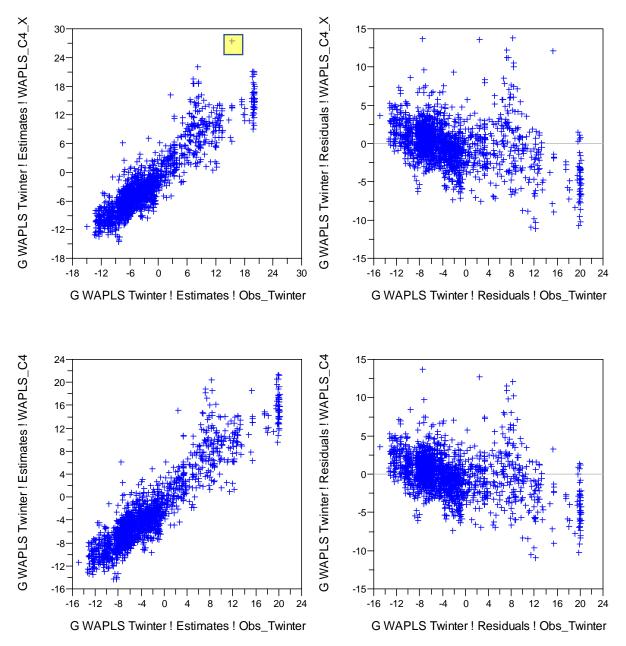
H. Observed graphs WA and WA-PLS WA Winter temperature Inverse deshrinking The yellow/blue boxes are the outliers

WA Total precipitation inverse deshrinking

The yellow/blue boxes are the outliers

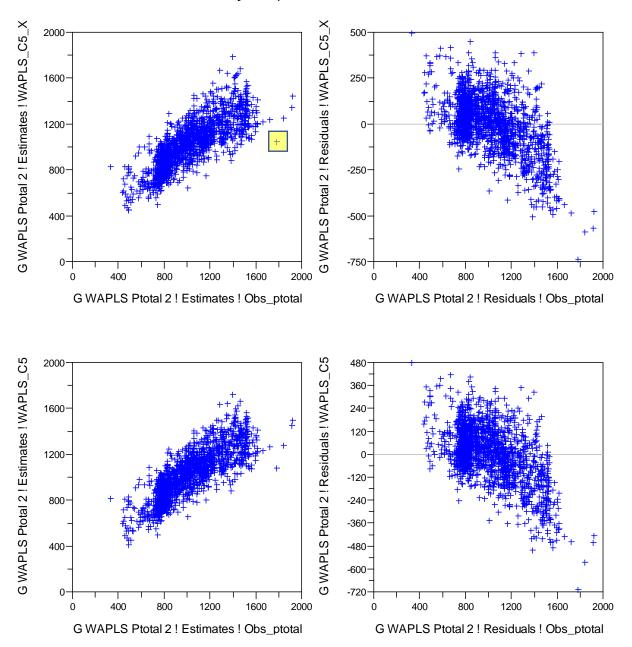


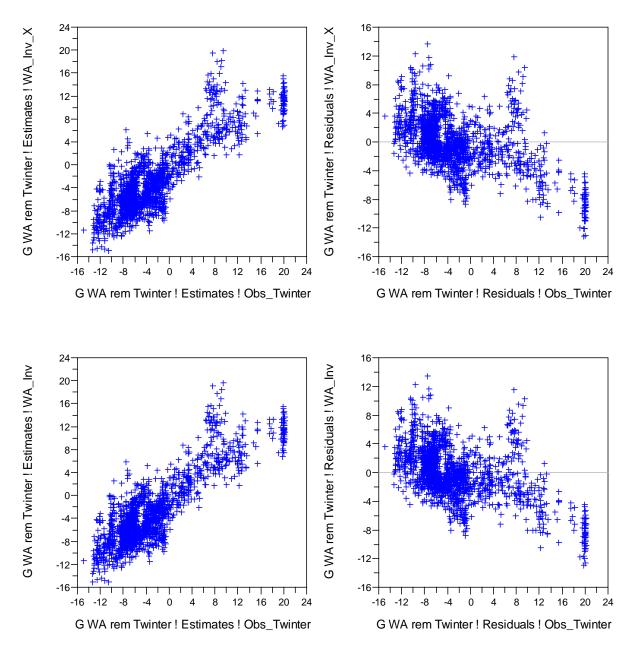
WAPLS Winter temperature C4 *The yellow/blue boxes are the outliers*



WAPLS Total precipitation C5

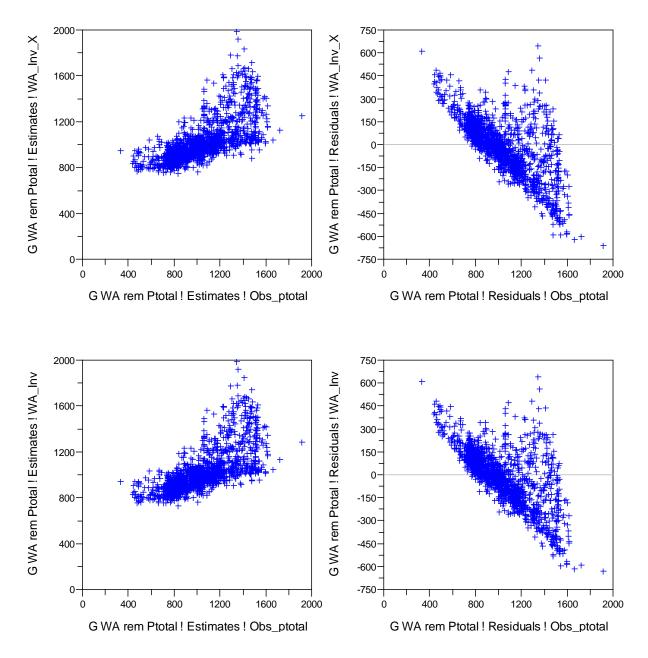
The yellow/blue boxes are outliers



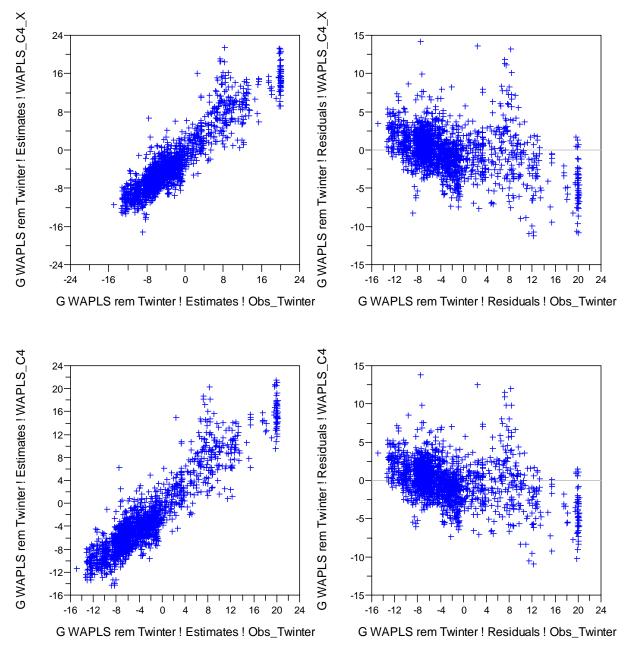


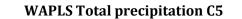
I. Observed graphs WA and WAPLS with removed samples WA Winter temperature inverse deshrinking

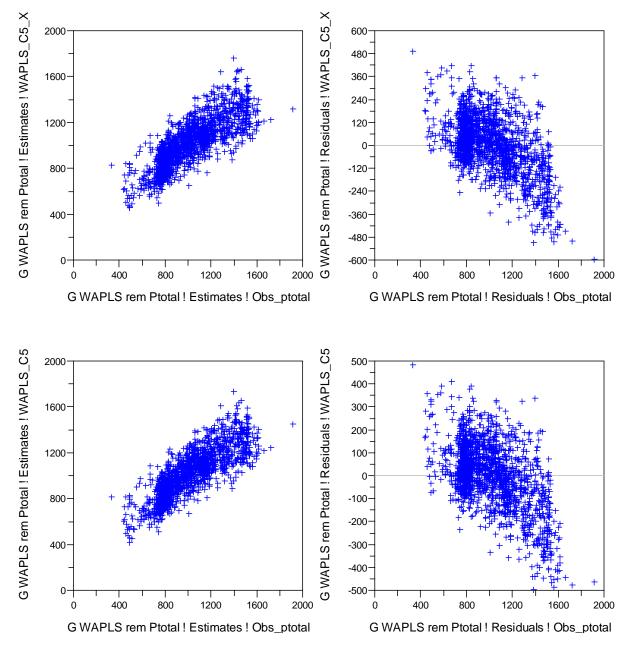
WA Total precipitation inverse deshrinking

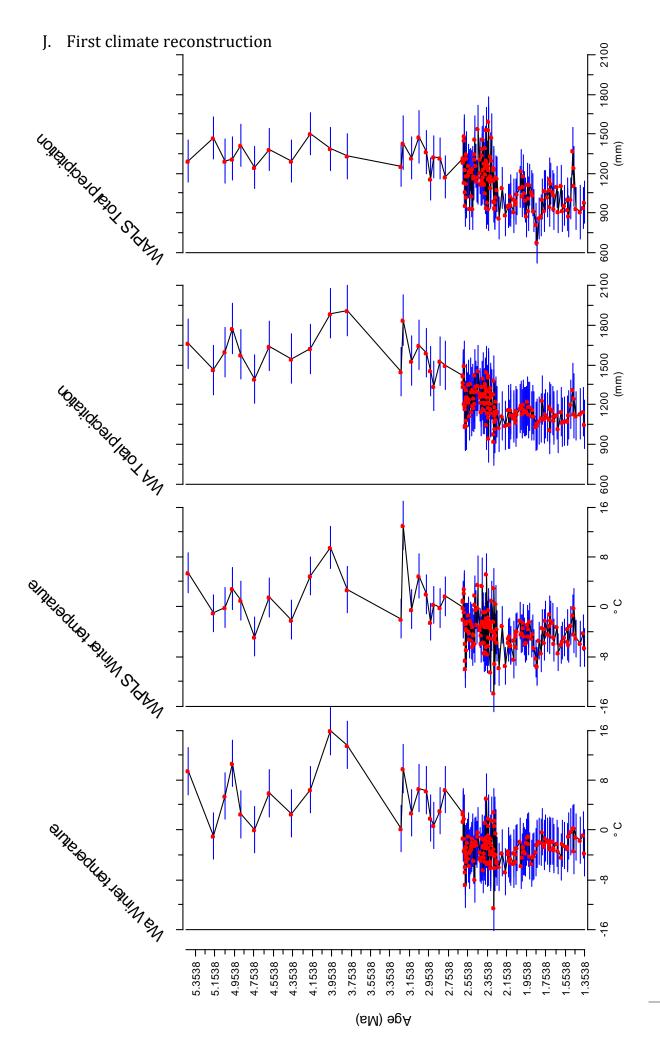


WAPLS Winter temperature C4









K. Calibration Sample number – age

	ration San	ipie in		je			
Sample nr.	Age (Ma)		Age (Ma)	07	Age (Ma)	100	0.40000
1	2.116098	44	2.424891	87	1.57887	130	2.13393
2	2.119061	45	2.431959	88	1.59743	131	2.16844
3	2.130577	46	2.443502	89	1.62904	132	2.19599
4	2.255519	47	2.450815	90	1.6331	133	2.2276
5	2.258005	48	2.456692	91	1.67109	134	2.26124
6	2.260425	49	2.474284	92	1.67138	135	2.28676
7	2.272114	50	2.482949	93	1.68414	136	2.30387
8	2.273652	51	2.485803	94	1.70067	137	2.3146
9	2.275655	52	2.488636	95	1.71326	138	2.32127
10	2.277604	53	2.49423	96	1.714505	139	2.34128
11	2.279502	54	2.500189	97	1.71575	140	2.35317
12	2.284025	55	2.505537	98	1.7382	141	2.36187
13	2.285751	56	2.51074	99	1.75635	142	2.37335
14	2.28743	57	2.513913	100	1.76418	143	2.37666
15	2.312246	58	2.519455	101	1.76447	144	2.38072
16	2.31425	59	2.530914	102	1.78622	145	2.40798
17	2.316169	60	2.541043	103	1.79753	146	2.48143
18	2.321493	61	2.549707	104	1.82351	147	2.53071
19	2.322326	62	2.556837	105	1.84074	148	2.58182
20	2.323146	63	2.562435	106	1.84219	149	2.58788
21	2.334567	64	2.56659	107	1.85031	150	2.59091
22	2.335763	65	2.569487	108	1.86887	151	2.59576
23	2.337516	66	2.571419	109	1.88192	152	2.60242
24	2.338917	67	2.572799	110	1.89729	153	2.78867
25	2.342445	68	2.574169	111	1.91846	154	2.84333
26	2.343159	69	2.576215	112	1.92455	155	2.90471
27	2.346052	70	2.579773	113	1.9318	156	2.93157
28	2.349784	71	2.585846	114	1.93615	157	2.975
29	2.352887	72	2.595611	115	1.95152	158	3.05091
30	2.360342	73	2.605367	116	1.95674	159	3.13
31	2.363895	74	2.615122	117	1.97501	160	3.21691
32	2.367609	75	1.35383	118	1.97704	161	3.23709
33	2.372486	76	1.36978	119	1.98139	162	3.79
34	2.377632	77	1.39356	120	1.99038	163	3.97
35	2.380848	78	1.43851	121	2.00865	164	4.17
36	2.384162	79	1.4591	121	2.01329	165	4.37
30	2.388728	80	1.46606	122	2.04345	165	4.59
38	2.389895	81	1.47331	123	2.05331	160	4.75
30	2.307073	82	1.4939	121	2.05534	167	4.89
40	2.400854	83	1.51159	125	2.06056	160	4.97
40	2.40597	84	1.5229	120	2.00030	107	5.05
41	2.40397	85	1.52464	127	2.08113	170	5.17
42	2.412339	86	1.56205	120	2.08321	171	5.43
43	2.417342	00	1.50205	129	2.11392	1/2	3.43

L. Fossil pollen species combined with modern pollen spec

Hank	Noordwijk	Petten	Whitemore
Abies	Abies	Abies	Abies
-	Acer	Acer	Acer +
Alisma	-	-	Not used
Alnus	Alnus	Alnus	Alnus +
Armeria	-	-	Armeria
Artemisia	Artemisia	Artemisia	Artemisia
-	-	Asteraceae	Aster
-	Azolla	-	Not used
Betula	Betula	Betula	Betula
-	Brassicaceae	-	Brassicaceae
Calluna	-	-	Ericaceae
Carpinus	Carpinus	Carpinus	Ostrycar
Carya	Carya	Carya	Carya
Caryophyllaceae	Caryophyllaceae	Caryophyllaceae	Caryophyllaceae
	Castanea	-	Castanea
Cedrus	Cedrus	-	Larix
Celtis	-	-	Celtis
	Cerebropollenites	-	Not used
Chenopodiaceae	Chenopodiaceae	Chenopodiaceae	Chenopodiaceae
-	Cicatricosisporites	-	Not used
-	Classopollis spp	-	Not used
C. Tubulifloraea	C. Tubulifloraea	-	Not used
C. Liguliforaea	C. Ligulifloraea	-	Not used
Corylus	Corylus	Corylus	Corylus
-	Cupressaceae	-	Cupressaceae
Cyperaceae	Cyperaceae	Cyperaceae	Cyperaceae
-	Dipsacaceae	-	Not used
Dryopteris	Dryopteris	Dryopteris	Not used
-	Elaeagnus	-	Elaegnaceae
-	Engelhardia	-	Not used
-	Ephedra	-	Ephedra
Equisetum	-	-	Equisetum
Ericales	Ericales	Ericales	Ericales
Eucommia	Eucommia	-	Not used
Euphorbia	-	-	Euphorbiaceae
Fagus	Fagus	Fagus	Fagus
Filipendula	-	-	Not used
Fraxinus	Fraxinus	Fraxinus	Fraxinus +
-	Gramineae	-	Not used
Hedera	Hedera	-	Not used
-	Illex	-	Aquifolicaeae
-	Isoëtes	-	Not used
Juglans	-	Juglans	Juglans +
-	-	Juniperus	Not used
-	-	Liliaceae	Liliaceae
Liquidambar	Liquidambar	-	Liquidambar
Lonicera	-	-	Not used
Lycopodium (exoot)	Lycopodium (exoot)	Lycopodium (exoot)	Lycopodium +

-	Lythrum	-	Not used
-	-	Malvaceae	Malvaceae
Myrica	Myricaceae	-	Myrica
-	Myriophyllum	-	Not used
-	Nymphaea	Nympheae	Not used
Nyssa	Nyssa	Nyssa	Nyssa
Osmunda	Osmunda	Osmunda	Osmundaceae
Ostrya	-	Ostrya	Ostryacar
Parthenociccus	-	-	Not used
-	-	Pilularia	Not used
Picea	Picea	Picea	Picea +
Pinus	Pinus	Pinus	Pinus
Pinus haplo	Pinus haplo	Pinus haplo	Pinus haplo
-	Plantago	-	Plantaginaceae
-	Plantago spp	-	Not used
-	Platycarya	-	Not used
Poaceae	-	Poaceae	Poaceae
Polypodium	_	-	Polypodiaceae
-	Potamogeton	-	Not used
Pteridium	Pteridium	Pteridium	Pteridium
Pterocarya	Pterocarya	Pterocarya	Not used
Quercus	Quercus	Quercus	Quercus
Ranunculus	Ranunculaceae	-	Ranunculaceae
-	Reveesia	-	Not used
-	Rhus	-	Not used
-	Rosaceae	-	Rosaceae
Rumex	Rumex		Rumex
-	Salix	Salix	Salix
Sanguisorba	-	-	Sanquisorba
-	Sapotaceae	_	Not used
Sciadopitys	Sciadopitys	-	Not used
Sequoia	Sequoia		Not used
-	Spergula	-	Caryophyllaceae
Sphagnum	Sphagnum	Sphagnum	Sphagnum
-	-	Succisa	Caprifoliaceae
-	Symplocos	-	Not used
Taxodium	Taxodium	Taxodium	Taxodium
Taxus	-	-	Taxus
-	-	- Thalictrum	Thalictrum
- Thelypteris	-	-	Not used
Tilia	- Tilia	- Tilia	Tilia
1 1110	Trudopollis	11110	Not used
Tsuga	Tsuga	Tsuga	Tsuga +
	Typhaceae	-	Not used
- Ulmus	Ulmus	- Ulmus	Ulmus
UIIIUS	Umbelliferaea	UIIIUS	
- Urtica	Univeriljeraea	- Urtica	Not used
	- Viscum		Urticaceae
-		-	Arceuthobium
-	Vitis	-	Not used

Trees	English name	Bisaccates	English name
Acer	Japanese maple	Abies	Firs
Alnus	Alder	Cupressa	Conifer
Anacardia	Cashew family	Picea	Spruce
Arecaceae	Palm	Pinus	Pine
Betula	Birch	Pinus Haploxyon	Pine
Carpinus	Hormbeam	Tsuga	Hemlock tree
Carya	Hickory		
Celtis	Nettle tree		
Corylus	Hazel		
Fagus	Beech		
Juglans	Walnut tree		
Fraxinus	Ash		
Liquidambar	American Storax		
Nyssa	Tupelo		
Quercus	Oak		
Salix	Willow		
Taxodium	Taxus		
Tilia	Lime tree		
Ulmus	Elm		

M. Percentage Trees & Bisaccates

Sample nr	Sum	Sum	Pollensum	%Trees	%Bisaccates
-	trees	Bisaccates			
1	10	12	22	45,45455	54,54545
2	14	2	16	87,5	12,5
3	52	23	75	69,33333	30,66667
4	5	23	28	17,85714	82,14286
5	39	103	142	27,46479	72,53521
6	43	113	156	27,5641	72,4359
7	107	55	162	66,04938	33,95062
8	116	39	155	74,83871	25,16129
9	50	87	137	36,49635	63,50365
10	32	113	145	22,06897	77,93103
11	70	72	142	49,29577	50,70423
12	71	76	147	48,29932	51,70068
13	76	85	161	47,20497	52,79503
14	74	82	156	47,4359	52,5641
15	38	138	176	21,59091	78,40909
16	28	111	139	20,14388	79,85612
17	17	119	136	12,5	87,5
18	59	89	148	39,86486	60,13514
19	51	82	133	38,34586	61,65414
20	62	52	114	54,38596	45,61404
21	40	75	115	34,78261	65,21739
22	60	44	104	57,69231	42,30769

24 56 65 121 46,28099 53,7190 25 85 50 135 62,96296 37,0370 26 111 40 151 73,50933 26,4900 27 88 44 132 66,66667 33,3333 28 68 67 135 50,37037 49,6296 29 61 91 152 40,13158 59,8684 30 38 75 113 33,6232 66,37166 31 59 49 108 54,62963 45,3703 32 25 76 101 24,75248 75,24753 33 36 61 97 37,1134 62,8664 34 43 61 104 41,3615 55,04583 36 37 68 105 35,2381 64,7619 37 32 64 96 33,3333 66,6666 38 31 66 97<						
25 85 50 135 62,96296 37,0370 26 111 40 151 73,5093 26,4900 27 88 44 132 66,6667 33,333 28 68 67 135 50,3737 49,6296 29 61 91 152 40,13158 59,8684 30 38 75 113 33,62832 66,3716 31 59 49 108 54,62963 45,3703 32 25 76 101 24,75248 75,2475 33 36 61 97 37,1134 62,886 34 43 61 104 41,34615 58,6538 36 37 68 105 35,2381 64,7615 37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,0412 39 38 99 137						12,94118
26 111 40 151 73,50993 26,4900 27 88 44 132 66,6667 33,3333 28 68 67 135 50,37037 49,62963 29 61 91 152 40,1318 59,8684 30 38 75 113 33,62832 66,37163 31 59 49 108 54,62963 45,3703 32 25 76 101 24,75248 75,24753 33 36 61 97 37,1134 62,8663 34 43 61 104 41,3445 58,6538 35 49 60 109 44,95413 55,0458 36 37 68 105 35,2381 64,6666 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 72,2627 40 29 74 103						
27 88 44 132 66,66667 33,3333 28 68 67 135 50,37037 49,6296 29 61 91 152 40,13158 59,8684 30 38 75 113 33,62832 66,37166 31 59 49 108 54,62963 45,37033 32 25 76 101 24,75248 75,24753 33 36 61 97 37,1134 62,8666 34 43 61 104 41,34615 58,6588 35 49 60 109 44,95413 55,0458 36 37 68 105 35,2381 64,7619 37 32 64 96 33,3333 66,66666 38 31 66 97 31,95876 68,80112 39 38 99 137 27,73723 72,2627 40 29 74 1						37,03704
28 68 67 135 50,37037 49,6296 29 61 91 152 40,13158 59,8684 30 38 75 113 33,62832 66,3716 31 59 49 108 54,62963 45,3703 32 25 76 101 24,75248 75,2475 33 36 61 97 37,1134 62,8866 34 43 61 104 41,34615 58,6538 36 37 68 105 35,2381 64,761 37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,8843 66,11535 42 69 11 37,8784						26,49007
29 61 91 152 40,13158 59,86843 30 38 75 113 33,62832 66,37163 31 59 49 108 54,62963 45,3703 32 25 76 101 24,75248 75,24753 33 36 61 97 37,1134 62,8864 34 43 61 104 41,34615 58,6538 36 37 68 105 35,2381 64,7649 37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,84466 41 41 80 121 33,8843 66,153 42 36 91 127 28,34646 71,6535 43 16 68 84 </th <th></th> <th></th> <th></th> <th></th> <th></th> <th></th>						
30 38 75 113 33,62832 66,37160 31 59 49 108 54,62963 45,3703 32 25 76 101 24,75248 75,2475 33 36 61 97 37,1134 62,868 34 43 61 104 41,34615 58,6538 36 37 68 105 35,2381 64,7619 37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,8843 66,115 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,9523 44 19 41 60						49,62963
31 59 49 108 54,62963 45,3703 32 25 76 101 24,75248 75,2475 33 36 61 97 37,1134 62,886 34 43 61 104 41,34615 58,6538 35 49 60 109 44,95413 55,0458 36 37 68 105 35,2381 64,7619 37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,8843 66,115 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,9523 44 19 41 60						59,86842
32 25 76 101 24,75248 75,24753 33 36 61 97 37,1134 62,8864 34 43 61 104 41,34615 58,6538 35 49 60 109 44,95413 55,0458 36 37 68 105 35,2381 64,6666 37 32 64 96 33,3333 66,66667 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,84466 41 41 80 121 33,843 66,115 42 36 91 127 28,34646 71,653-4 43 16 68 84 19,04762 28,9333 44 19 41 60 31,86869 86,1111 44 19 41 60						66,37168
33 36 61 97 37,1134 62,886 34 43 61 104 41,34615 58,6538 35 49 60 109 44,95413 55,0458 36 37 68 105 35,2381 64,761 37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 37,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,8843 66,115 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,9523 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 56 27 77 104						
34 43 61 104 41,34615 58,6538 35 49 60 109 44,95413 55,0458 36 37 68 105 35,2381 64,7619 37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,8843 66,115 42 36 91 127 28,34646 71,65354 43 16 68 84 19,04762 80,9523 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95						
35 49 60 109 44,95413 55,04583 36 37 68 105 35,2381 64,7619 37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,0412 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,843 66,157 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,9523 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,795 47 18 77 95 18,94737 81,0526 44 169 111 37,83784 <th></th> <th></th> <th></th> <th></th> <th></th> <th>62,8866</th>						62,8866
36 37 68 105 35,2381 64,7619 37 32 64 96 33,3333 66,6667 38 31 66 97 31,95876 68,04124 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,84466 41 41 80 121 33,8843 66,1157 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,9523 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95 18,94737 81,05263 48 25 69 94 26,59574 73,40424 49 25 129 154 <th></th> <th></th> <th></th> <th></th> <th></th> <th>58,65385</th>						58,65385
37 32 64 96 33,3333 66,6666 38 31 66 97 31,95876 68,04124 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,8843 66,1157 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,95233 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95 18,94737 81,0526 48 25 69 94 26,59574 73,40424 49 25 129 154 16,23377 83,7623 50 27 77 104 <th></th> <th></th> <th></th> <th></th> <th></th> <th>55,04587</th>						55,04587
38 31 66 97 31,95876 68,04124 39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,8843 66,1157 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,95233 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95 18,94737 81,0526 48 25 69 94 26,59574 73,4042 49 25 129 154 16,23377 83,76623 50 27 77 104 25,96154 74,0384 51 42 69 111 </th <th></th> <th></th> <th></th> <th></th> <th></th> <th>64,7619</th>						64,7619
39 38 99 137 27,73723 72,2627 40 29 74 103 28,15534 71,8446 41 41 80 121 33,843 66,115 42 36 91 127 28,34646 71,65354 43 16 68 84 19,04762 80,95236 44 19 41 60 31,66667 68,33333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95 18,94737 81,05263 48 25 69 94 26,59574 73,4042 49 25 129 154 16,23377 83,76623 50 27 77 104 25,96154 74,03844 51 42 69 111 37,83784 62,16216 53 40 67 10			64			
40 29 74 103 28,15534 71,8446 41 41 80 121 33,8843 66,115 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,95233 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95 18,94737 81,05265 48 25 69 94 26,59574 73,40420 49 25 129 154 16,23377 83,76623 50 27 77 104 25,96154 74,03844 51 42 69 111 37,83784 62,16210 52 38 72 110 34,54545 65,45453 53 40 67 9						68,04124
41 41 80 121 33,8843 66,115 42 36 91 127 28,34646 71,6535 43 16 68 84 19,04762 80,95236 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95 18,94737 81,05265 48 25 69 94 26,59574 73,40420 49 25 129 154 16,23377 83,76623 50 27 77 104 25,96154 74,03840 51 42 69 111 37,83784 62,16210 52 38 72 110 34,54545 65,45453 53 40 67 107 37,38318 62,61682 54 37 51 88 42,04545 57,95453 55 28 67 95 <	39				27,73723	72,26277
42 36 91 127 28,34646 71,65354 43 16 68 84 19,04762 80,95233 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7953 47 18 77 95 18,94737 81,05263 48 25 69 94 26,59574 73,40420 49 25 129 154 16,23377 83,76623 50 27 77 104 25,96154 74,03840 51 42 69 111 37,83784 62,16210 52 38 72 110 34,54545 65,45453 53 40 67 107 37,38318 62,61682 53 40 67 95 29,47368 70,52633 55 28 67 <					28,15534	71,84466
43 16 68 84 19,04762 80,95233 44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95 18,94737 81,05263 48 25 69 94 26,59574 73,40420 49 25 129 154 16,23377 83,76623 50 27 77 104 25,96154 74,0384 616 67 107 37,83784 62,16210 51 42 69 111 37,83784 62,16210 52 38 72 110 34,54545 65,45455 53 40 67 95 29,47368 70,52633 55 28 67 95 29,47368 70,52633 55 28 67 95 <t< th=""><th>41</th><th>41</th><th>80</th><th>121</th><th>33,8843</th><th>66,1157</th></t<>	41	41	80	121	33,8843	66,1157
44 19 41 60 31,66667 68,3333 45 15 93 108 13,8889 86,1111 46 16 77 93 17,2043 82,7957 47 18 77 95 18,94737 81,05263 48 25 69 94 26,59574 73,40420 49 25 129 154 16,23377 83,76623 50 27 77 104 25,96154 74,03840 51 42 69 111 37,83784 62,16210 52 38 72 110 34,54545 65,45453 53 40 67 107 37,38318 62,61683 54 37 51 88 42,04545 57,95453 55 28 67 95 29,47368 70,52633 55 28 67 95 29,47368 70,52633 56 29 36 <t< th=""><th></th><th>36</th><th>91</th><th>127</th><th>28,34646</th><th>71,65354</th></t<>		36	91	127	28,34646	71,65354
45159310813,888986,11114616779317,204382,7954718779518,9473781,05264825699426,5957473,40426492512915416,2337783,766250277710425,9615474,0384651426911137,8378462,1621652387211034,5454565,4545553406710737,3831862,61685437518842,0454557,954555528679529,4736870,526335629366544,6153855,384665718415930,5084769,4915558349212626,9841373,015876032508239,0243960,9756761268811422,8070277,19296623011014021,4285778,5714564199111017,2727382,7272	43	16	68	84	19,04762	80,95238
4616779317,204382,79574718779518,9473781,052634825699426,5957473,40420492512915416,2337783,7662350277710425,9615474,0384051426911137,8378462,1621052387211034,5454565,4545353406710737,3831862,616835437518842,0454557,954535528679529,4736870,526335629366544,6153855,384635718415930,5084769,4915358349212626,9841373,015836032508239,0243960,9756361268811422,8070277,19293623011014021,4285778,5714364199111017,2727382,7272	44	19	41	60	31,66667	68,33333
47 18 77 95 18,94737 81,05263 48 25 69 94 26,59574 73,40426 49 25 129 154 16,23377 83,76623 50 27 77 104 25,96154 74,03846 51 42 69 111 37,83784 62,16216 52 38 72 110 34,54545 65,45453 53 40 67 107 37,38318 62,61683 54 37 51 88 42,04545 57,95453 55 28 67 95 29,47368 70,52633 55 28 67 95 29,47368 70,52633 56 29 36 65 44,61538 55,38463 57 18 41 59 30,50847 69,49153 58 34 92 126 26,98413 73,01583 58 34 92	45	15	93	108	13,88889	86,11111
4825699426,5957473,40420492512915416,2337783,7662350277710425,9615474,0384051426911137,8378462,1621052387211034,5454565,4545353406710737,3831862,616835437518842,0454557,954535528679529,4736870,526335629366544,6153855,384635718415930,5084769,4915358349212626,9841373,015836032508239,0243960,9756361268811422,8070277,19293623011014021,4285778,57143632711714418,7581,2364199111017,2727382,7272	46	16	77	93	17,2043	82,7957
492512915416,2337783,7662350277710425,9615474,0384651426911137,8378462,1621652387211034,5454565,4545353406710737,3831862,616835437518842,0454557,954535528679529,4736870,526335629366544,6153855,384635718415930,5084769,4915358349212626,9841373,015836032508239,0243960,9756361268811422,8070277,19293623011014021,4285778,57143632711714418,7581,2364199111017,2727382,7272	47	18	77	95	18,94737	81,05263
50 27 77 104 25,96154 74,03844 51 42 69 111 37,83784 62,16216 52 38 72 110 34,54545 65,45455 53 40 67 107 37,38318 62,6168 54 37 51 88 42,04545 57,95455 55 28 67 95 29,47368 70,52637 56 29 36 65 44,61538 55,38467 57 18 41 59 30,50847 69,49157 58 34 92 126 26,98413 73,01587 59 63 72 135 46,66667 53,33333 60 32 50 82 39,02439 60,97567 61 26 88 114 22,80702 77,19298 62 30 110 140 21,42857 78,57143 63 27 117	48	25	69	94	26,59574	73,40426
51426911137,8378462,1621652387211034,5454565,454553406710737,3831862,61685437518842,0454557,95455528679529,4736870,52635629366544,6153855,38465718415930,5084769,491558349212626,9841373,015859637213546,6666753,33336032508239,0243960,975661268811422,8070277,1929623011014021,4285778,57143632711714418,7581,2364199111017,2727382,7272	49	25	129	154	16,23377	83,76623
52387211034,5454565,4545353406710737,3831862,616835437518842,0454557,954535528679529,4736870,526335629366544,6153855,384635718415930,5084769,4915358349212626,9841373,0158359637213546,6666753,33336032508239,0243960,9756361268811422,8070277,19293623011014021,4285778,5714364199111017,2727382,72723	50	27	77	104	25,96154	74,03846
53406710737,3831862,616825437518842,0454557,954535528679529,4736870,526335629366544,6153855,384635718415930,5084769,4915358349212626,9841373,0158359637213546,6666753,33336032508239,0243960,9756361268811422,8070277,19298623011014021,4285778,57143632711714418,7581,2364199111017,2727382,7272	51	42	69	111	37,83784	62,16216
5437518842,0454557,954535528679529,4736870,526335629366544,6153855,384635718415930,5084769,4915358349212626,9841373,0158359637213546,6666753,33336032508239,0243960,9756361268811422,8070277,19298623011014021,4285778,5714364199111017,2727382,72723	52	38	72	110	34,54545	65,45455
5528679529,4736870,526325629366544,6153855,384625718415930,5084769,4915358349212626,9841373,0158359637213546,6666753,33336032508239,0243960,9756361268811422,8070277,19298623011014021,4285781,2364199111017,2727382,72723	53	40	67	107	37,38318	62,61682
5629366544,6153855,384625718415930,5084769,4915358349212626,9841373,0158359637213546,6666753,333336032508239,0243960,9756361268811422,8070277,19298623011014021,4285778,57143632711714418,7581,2364199111017,2727382,7272	54	37	51	88	42,04545	57,95455
5718415930,5084769,4915358349212626,9841373,0158359637213546,6666753,33336032508239,0243960,9756361268811422,8070277,19298623011014021,4285778,57143632711714418,7581,2364199111017,2727382,7272	55	28	67	95	29,47368	70,52632
58 34 92 126 26,98413 73,0158 59 63 72 135 46,66667 53,3333 60 32 50 82 39,02439 60,9756 61 26 88 114 22,80702 77,19298 62 30 110 140 21,42857 78,57143 63 27 117 144 18,75 81,25 64 19 91 110 17,27273 82,7272	56	29	36	65	44,61538	55,38462
59637213546,6666753,33336032508239,0243960,9756361268811422,8070277,19298623011014021,4285778,57143632711714418,7581,2364199111017,2727382,72723	57	18	41	59	30,50847	69,49153
6032508239,0243960,975661268811422,8070277,19298623011014021,4285778,57143632711714418,7581,2964199111017,2727382,72723	58	34	92	126	26,98413	73,01587
61268811422,8070277,19298623011014021,4285778,57143632711714418,7581,2364199111017,2727382,72723	59	63	72	135	46,66667	53,33333
623011014021,4285778,57143632711714418,7581,2364199111017,2727382,72723	60	32	50	82	39,02439	60,97561
63 27 117 144 18,75 81,25 64 19 91 110 17,27273 82,72727	61	26	88	114	22,80702	77,19298
64 19 91 110 17,27273 82,72727	62	30	110	140	21,42857	78,57143
	63	27	117	144	18,75	81,25
65 19 98 117 16 23932 83 76069	64	19	91	110	17,27273	82,72727
11 10 11 10 10 10 10	65	19	98	117	16,23932	83,76068
66 27 114 141 19,14894 80,85106	66	27	114	141	19,14894	80,85106
67 39 156 195 20 80	67	39	156	195	20	80

68	26	116	142	18,30986	81,69014
69	28	84	112	25	75
70	7	177	184	3,804348	96,19565
71	78	96	174	44,82759	55,17241
72	74	110	184	40,21739	59,78261
73	25	175	200	12,5	87,5
74	1	1	2	50	50
75	1	6	7	14,28571	85,71429
76	5	2	7	71,42857	28,57143
77	1	3	4	25	75
78	1	0	1	100	0
79	3	16	19	15,78947	84,21053
80	8	83	91	8,791209	91,20879
81	4	120	124	3,225806	96,77419
82	3	64	67	4,477612	95,52239
83	3	0	3	100	0
84	13	39	52	25	75
85	0	1	1	0	100
86	2	0	2	100	0
87	1	4	5	20	80
88	9	1	10	90	10
89	0	3	3	0	100
90	6	106	112	5,357143	94,64286
91	37	26	63	58,73016	41,26984
92	20	31	51	39,21569	60,78431
93	7	6	13	56	44
94	17	65	82	20,73171	79,26829
95	22	2	24	93,61702	6,382979
96	15	19	34	44,11765	55,88235
97	2	1	3	66,66667	33,33333
98	7	3	10	70	30
99	12	79	91	13,18681	86,81319
100	19	57	76	25	75
101	12	76	88	13,63636	86,36364
102	13	62	75	17,33333	82,66667
103	9	14	23	39,13043	60,86957
104	3	5	8	37,5	62,5
105	2	8	10	20	80
106	6	7	13	46,15385	53,84615
107	3	5	8	37,5	62,5
108	13	62	75	17,33333	82,66667
109	8	41	49	16,32653	83,67347
110	13	26	39	33,33333	66,66667
111	8	82	90	8,888889	91,11111
112	9	49	58	15,51724	84,48276

113	8	13	21	38,09524	61,90476
113	6	37	43	13,95349	86,04651
114	23	64	87	26,5896	73,4104
115	6	52	58	10,34483	89,65517
110	9	30	39	23,07692	76,92308
117	4	13	17	23,52941	76,47059
110	11	23	34	32,35294	67,64706
120	14	44	58	24,13793	75,86207
120	8	4	12	66,66667	33,33333
121	4	19	23	17,77778	82,22222
122	3	16	19	15,78947	84,21053
123	3	39	42	7,142857	92,85714
121	14	60	74	19,04762	80,95238
125	2	38	40	5	95
120	9	59	68	13,33333	86,66667
127	16	75	91	17,58242	82,41758
120	8	75	83	9,638554	90,36145
130	15	43	58	25,86207	74,13793
131	5	94	99	5,050505	94,94949
132	10	15	25	40	60
133	5	71	76	6,578947	93,42105
134	8	90	98	8,163265	91,83673
135	2	68	70	2,857143	97,14286
136	16	59	75	21,33333	78,66667
137	5	7	12	41,66667	58,33333
138	5	69	74	6,756757	93,24324
139	9	47	56	16,07143	83,92857
140	2	21	23	8,695652	91,30435
141	41	37	78	52,90323	47,09677
142	17	76	93	18,37838	81,62162
143	11	77	88	12,57143	87,42857
144	16	53	69	23,35766	76,64234
145	7	76	83	8,433735	91,56627
146	15	125	140	10,71429	89,28571
147	8	164	172	4,651163	95,34884
148	16	86	102	15,68627	84,31373
149	9	22	31	29,03226	70,96774
150	8	36	44	18,18182	81,81818
151	14	183	197	7,106599	92,8934
152	27	128	155	17,41935	82,58065
153	34	65	99	34,34343	65,65657
154	8	39	47	17,02128	82,97872
155	18	38	56	32,14286	67,85714
156	19	135	154	12,33766	87,66234
157	23	99	122	18,85246	81,14754

158	60	293	353	16,99717	83,00283
159	74	341	415	17,83133	82,16867
160	15	124	139	10,79137	89,20863
161	20	191	211	9,478673	90,52133
162	140	375	515	27,18447	72,81553
163	35	43	78	44,87179	55,12821
164	57	166	223	25,56054	74,43946
165	13	118	131	9,923664	90,07634
166	22	232	254	8,661417	91,33858
167	19	99	118	16,10169	83,89831
168	30	495	525	5,714286	94,28571
169	9	20	29	31,03448	68,96552
170	20	77	97	20,61856	79,38144
171	11	175	186	5,913978	94,08602
172	28	102	130	21,53846	78,46154