1	PhylochipAnalyzer - A Program for Analysing Hierarchical Probe-Sets
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# 1 Abstract

2 The recent introduction of phylochips that contain molecular probes facilitates environmental 3 microbial identification in a single experiment without previous cultivation. A set of probes 4 recognizing species at different taxonomic levels is denoted as a hierarchical set. Application 5 of hierarchical probe sets on a DNA-microarray allows the assessment of biodiversity with 6 different resolutions. It significantly increases the robustness of the results retrieved from 7 phylochip experiments because of the possible consistency checks of hybridization across 8 different taxonomic levels. Here, we present a computer program, PhylochipAnalyzer, for the 9 hierarchy editing and the evaluation of phylochip data generated from hierarchical probe-sets.

1 Basic rationale: Recently, more and more publications describe the application of DNA 2 microarrays for species identification (phylochips) from environmental samples (Guschin et 3 al. 1997; Loy et al. 2002; Metfies and Medlin 2004; Call 2005, Medlin et al. 2006). 4 Phylochips are DNA-microarrays containing molecular probes that bind to unique sequences 5 in a target. The target sequence is usually part of marker genes, e.g., the ribosomal RNA gene. 6 Ribosomal RNA-genes are particularly well suited for phylochip- and phylogenetic analysis, 7 because they are universal, found in all cellular organisms, are of relatively large size; and 8 contain both highly conserved and variable regions with no evidence for lateral gene transfer 9 (Woese 1987). The large number of published 18S rDNA-sequences, (e.g., RDP, Maidak et 10 al. 2001) makes it possible to design hierarchical probe sets that specifically target the 18S-11 rDNA from higher taxa down to species level (Lange et al. 1996; Guillou et al. 1999; Groben 12 et al. 2004). Phylochips provide a promising tool to identify large numbers of microbial 13 species in complex environmental samples quickly without a cultivation step. Our phylochip 14 contains a hierarchical set of probes, which target phytoplankton species at different 15 taxonomic levels (Metfies and Medlin 2004; Medlin et al. 2006). In a hierarchical probe-set, 16 a target species is only considered present, if all hierarchical probes for each species result in 17 a positive signal. Therefore, hierarchical probes add to the accuracy of molecular probe based 18 identification approaches.

In spite of the growing number of applications for phylochips, they represent only a
small proportion of all DNA-microarray related work. Most publications describe expression
studies (e.g., Lockhart et al. 2000; Stoughton 2005; Rensink 2005; Csako 2006).

Consequently, the majority of protocols are optimized for applications related to expression
analysis. However, the application of phylochips for species identification in environmental
samples presents technical challenges that are not encountered in gene expression studies of
laboratory samples (Peplies et al. 2003; Call et al. 2005; Metfies et al. 2006). There are

1 numerous commercial and non-commercial programs for the analysis of expression studies (e.g. Dondrup et al. 2003; Vaquerizas et al. 2005) but few programs exist for phylochip 2 3 analysis. One example is the Unix-based program ChipChecker (Loy et al. 2002), which is 4 dedicated to data interpretation from phylochips. It calibrates signal to noise ratios to a set 5 threshold determined by the user and finds positive signals with respect to that threshold 6 based on the fact that a positive signal can only be located where there is a fully 7 complementary probe to its target. However, in a hierarchical probe set, a signal is only 8 considered truly positive, if all probes in the hierarchy are positive. Therefore, the analysis of 9 hierarchically organized phylochips requires an additional step in comparison to the functions 10 provided by ChipChecker. The positive signals must be tested for their robustness in relation 11 to the hierarchy on the phylochip. In summary, a program for the analysis of hierarchically 12 organised phylochips has to provide an algorithm for the calculation of a signal to noise-value 13 and a tool that allows to set positive signals in relation to the hierarchy inherent in the design 14 of the probe-set. Here we present the program, PhylochipAnalyzer, that implements the 15 calculation of signal to noise ratios and the evaluation of phylochip-data with respect to probe 16 hierarchy.

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### 18 **Funcionality and Implementation aspects of the Program**

20 PhylochipAnalyzer is a GUI-based Windows-program, developed under Borland-Delphi. The 21 program combines two strongly interconnected functions: hierarchy editing and data analysis. 22 The user starts editing interactively and graphically the hierarchy that is inherent in the 23 chip/probe design process. Editing is started by loading a spot description file in GAL-format 24 generated by the GenePix- software (Axon Instruments Inc., USA). A procedure to convert 25 other formats is described in the software documentation. Spot entries are shifted manually so 26 that a hierarchically structured tree-like layout appears, in correspondence to the hierarchical 27 probe design of the chip seen in Fig. 2A, upper part. Probes must not be placed in a hierarchy

at all, e.g., positive or negative controls should be placed as stand-alone, i.e., with no parent
probes and no child probes. However, a positive control could be placed as the parent probe to
all other. The hierarchy is then saved as an XML-file that is used later for data analysis.
Whereas the XML-file stores the pure hierarchy information of the chip, spot-intensity data
are read from files with externally defined format, such as tab-delimited tables. The user may
include the probe sequence in the comment field. The hierarchy can be exported as a tree file
in Newick-format.

8 The second mode of operation is for the analysis of processed scanner data, i.e., tables 9 with data for foreground and background intensities of the individual spots. The presence or 10 absence of a hybridization signal is checked by a threshold criterion. The foreground-11 background intensity contrasts are normalized with respect to intensities of the negative 12 control spots (Loy et al. 2002). Here intensity data of multiple copies (blocks) of the spots on 13 each chip are evaluated and means and standard deviations are computed. The results for the 14 blocks on the chip are shown independently (Fig. 2A, bottom right) such that entire blocks 15 can be excluded from the analysis. It is assumed that if some spots in a certain block are 16 identified as outliers or if positive controls fail, the user should exclude the whole block from 17 evaluation because of the questionable quality of hybridization. A false positive signal on a 18 higher hierarchical level has consequences for the validity of lower levels, down to the 19 species level: PhylochipAnalyzer marks all positive signals that are below the hierarchy level 20 of a spot showing a negative signal, i.e., corrected lines are crossed out. Because a signal is 21 marked positive when the majority of copies give signals above the threshold, a correction is 22 always contradicting. The user should inspect whether the underlying probe is correctly 23 designed or maybe placed in the wrong hierarchy level.

The user may export the evaluation results directly to an Excel-graph (Fig.2B) in which the signals are given as bars, labelled with the probe identifier. The size of a bar indicates the

quality above the threshold, i.e., the longer the bar, the stronger the evidence for a positive
 signal. All data are shown with error bars of the mean due to the variance over the different
 blocks.

4 Validation

5 The PhylochipAnalyzer was used to analyse data retrieved from a hybridization of PCR-6 products of *Micromonas pusilla* 18S rDNA to a phylochip that contained 44 probes, including 7 a hierarchical probe-set for the Prasinophyte genus Micromonas. The hierarchical probe-set 8 consisted of six probes that bind, respectively, at the level of Kingdom (EUK 1209, EUK 9 328), Class (Chlo01, Chlo02), Clade or Order (Pras 04) and Genus (Micro01) to Micromonas 10 *pusilla*. The additional probes on the chip identified other phytoplankton taxa, a negative 11 control, and two positive controls. Fluorescence images of the hybridized phylochips were 12 taken with the Genepix 4000B Scanner (Axon Instruments Inc. USA). The signal intensities 13 were quantified using the GenePix 6.0 software (Axon Instruments Inc. USA). Raw data were 14 saved as a GPR-file and imported to the PhylochipAnalyzer-program. The computation of the 15 raw data with the PhylochipAnalyzer-program identified only positive signals for the 16 perfectly matching probes. For those probes, a signal/noise ratio was calculated that was 17 above the threshold. The complete hierarchical probe set resulted in positive signals, therefore 18 the signal for Micro01 can be considered truly positive (see Fig. 1 and Fig. 2B).

# 19 **Discussion**

The program simplifies tremendously the time consuming tasks of data processing of results from hierarchical phylochips. This is from particular interest, if high-throughput data are analyzed. The program is flexible with respect to configuration because the user can influence the threshold criterion by modifying the code that is implemented as a Delphi-script. This allows arbitrary modifications of the basic formula of data processing. Other formats of intensity description can easily be converted into appropriate GAL-format. On screen, the

1 user may change the threshold value (default 2) interactively for sensitivity studies and 2 recalculation. The rather simple criterion for elimination of false positives could be extended 3 towards more quantitative measures. We plan to extend the program for quantitative analysis, 4 i.e., spots from higher hierarchical levels are expected to show stronger signals than the lower 5 hierarchical spots because they target more individuals. Multi-chip comparative analysis (e.g., 6 clustering) for time-series analysis is also a desirable feature. The proposed XML-format for 7 hierarchy representation can be seen as a prototype for standardization in phylochip hierarchy 8 description. It is now necessary to introduce community standards for the representation of 9 both, chip description and data-processing details. For gene-expression analysis by means of 10 DNA-microarrays guidelines already exist (Brazma et al. 2001). Standards for phylochip 11 design and processing description are considered to be a prerequisite for permanent archiving 12 of publication supplemental data accompanied by catalogues of metadata in repositories.

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# 19 **Copyright, Download**

20 The copyright is specified by the author of the software (PB). The use of PhylochipAnalyzer

- 21 is free of charge. For software and supplemental material see
- 22 <u>http://www.awi.de/en/go/phylochipanalyzer</u>.

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**Figure 2:** 

### 2 A

	y-edit mode	Analyser	mode	Thresh	old val	ue: 3						
ID	ID		Name		1.0	Descripton	Mean valu	e stdm	Matching	g Spo		
EUK328     Chlo01     Chlo02     Chlo02		EUK328 Chlo01			All Eukaryotes	67,143	38,093	100%				
					Chlorophytes	18,351	11,06	100%				
		Chlo	02		Chlorophytes		8,8477	100%				
PRAS01			PRAS01 PRAS03 PRAS04			Pseudoscourfieldia marina,	1,1775	0,71934	100%			
PRAS03					Prasinococcales (Clade VI)	2,3802	1,4018	75%				
					Mamiellales (Clade II)	6,7943	4,2009	75%				
	MICR	1001	MICF	R001		Micromonas pusilla	4,0538	2,2849	100%			
OSTREO01			OST	REO01		Osterococcus tauri	1,9298	1,2005	75%	%		
	BATH	HY01	BATH	HY01		Bathycoccus prasinos	1,2594	0,87918	100%			
	PRAS05		PRA	505		Clade VIIA (CCMP 1205+RC	1,764	1,2861	75%			
PRAS06			PRA	506		Clade VIIB (environmental s	1,2693	0,79394	100%			
PRAS07			PRA	507		Pseudoscourfieldia marina,	2,6075	2,0681	67%			
	PRAS08		PRA	508		Picocystis salinarium (Clade	1,3534	0,79588	100%			
	PRAS094	41	PRA	509A1		Pterosperma cristatum (Clad	0,29265	0,21654	100%			
	PRAS094	42	PRA	509A2		Pyramimonas, Prasinopapilla	0,59475	0,43772	100%			
	PRAS09E	5	PRA	509D		Halosphaera spec. (Clade I) 📂	0,89999	0,55487	100%			
	PRAS10E	3	PRA	510B		Nephroselmis pyriformis, Ps	2,4289	1,3716	100%			
	PRAS10F	ŧ	PRA	510F		Nephroselmis pyriformis (diff	0,26924	0,24731	100%			
	PRAS10H	4	PRA	510H		Nephroselmis oliviaceae (Cla	1,0974	0,7146	100%			
<						- III				>		
Use i	Calcu	ID	F635	B635	Norr	n   Auto		Block Ma	tching spots			
	56,317	EUK328	3625	63	0	0		📝 1 🛛 46°	% (44/96)			
64,859 EUK328		EUK328	4630	71	0	0		2 439	% (41/96)			
	61,463	EUK328	4160	67	0	0		🗹 3 459	% (43/96)			
· · · · · ·	85,932	EUK328	6339	73	0	0 0		🗹 4 43°	% (41/96)			



**Figure 2. A:** Screenshot of the analyser-mode. Any set of molecular probes can be organized as a user defined phylogenetic tree by a drag and drop function in editor-mode. The screenshot displays a tree of probes that bind to Prasinophytes at different hierarchical levels. The bottom part shows an individual probe result for the selected probe (EUK328, top part). **B**: Output of signal-noise values in graphical form.

**Figure 1:** 



**Figure 1:** The 18S rDNA of *Micromonas pusilla* was hybridized to a set of 44 probes. The set of probes contained a hierarchical set that binds to the 18S rDNA of *M. pusilla* at four different taxonomic levels (EUK 1209, EUK 328, Chlo01, Chlo02, Pras04 and Micro01).