Predicting, exploring, and synthesising ancestral sequences using **Graphical Representation** of Ancestral Sequence Predictions (GRASP)

Gabe Foley

School of Chemistry and Molecular Biosciences The University of Queensland 10/12/2019

Overview

Ancestral Sequence Reconstruction (ASR)

- What is it?
- Why use it?
- Current restrictions on data set size

Graphical Representation of Ancestral Sequence Predictions (GRASP)

- Enables much greater data set sizes
- Valid ancestral predictions
- Allows for novel types of ancestors



What is ancestral sequence reconstruction?

• Using the information in modern day biological sequences to infer what their ancestors looked like



What is ancestral sequence reconstruction?

• Using the information in modern day biological sequences to infer what their ancestors looked like



What is ancestral sequence reconstruction?

 Using the information in modern day biological sequences to infer what their ancestors looked like

 Ancestral sequences can be 'resurrected' – synthesised and studied alongside modern day proteins



1. Collect sequences



2. Align sequences



3. Infer phylogenetic tree











5. Concatenate predictions into a complete sequence



PPTG PP-G PSAG PSAC S-TC S-SG PRVG -RTG



- Joint reconstruction all ancestors simultaneously
- Marginal reconstruction single ancestor



Why use ancestral sequence reconstruction?

- Studying evolutionary histories
- Determining important functional residues
- Utilising the ancestors for industrial applications



Why use ancestral sequence reconstruction?

- Studying evolutionary histories
- Determining important functional residues
- Utilising the ancestors for industrial applications



Why use ancestral sequence reconstruction?

- Studying evolutionary histories
- Determining important functional residues
- Utilising the ancestors for industrial applications



Adapted from Gumulya et al., Nature Catalysis 1, 878 (2018).

Current bottlenecks

- Garcia and Kaçar (2019) reviewed 12 ASR studies from the past decade
 - Data set sizes ranged from 21 to 456 sequences
 - Average of 168 sequences
- ASR tools such as FastML, PAML are capable of ~500-600 sequences
- GRASP is capable of ~10,000 sequences

GRASP data structure and implementation

- Partial order graphs
 - Represent ambiguity
 - Summarise insertion and deletion events
- Variable elimination
 - Decompose conditional probability tables into smallest number of operations

GR--P G-AS-G-ASP G-ASP



Validation – smaller data set sizes

We inferred a cytochrome P450 CYP2U1 ancestor (359 sequences) using GRASP, FastML, and PAML.

Regardless of the tool used, ancestral proteins -

- expressed at similar levels in E. coli
- displayed a P450 spectra
- had activity towards luciferin MultiCYP substrate
- showed similar thermal stabilities





Thermal stability of cytochrome P450 CYP2U1 ancestor



Validation – larger data set sizes

- Dihydroxy-acid dehydratase family (DHAD)
 a) DHAD phylogenetic trees of 1612 vs 9112 sequences
- As data set size increased ancestors were constrained towards canonical forms
- Ancestors from both the smallest and largest reconstructions were resurrected and showed activity towards D-Gluconate.

8 sequences 1389 sequences 97 sequences N9 1098 sequences N9 3983 387 sequences sequences N1087 3 sequences N1088 18 sequences 2806 sequences N1442 18 sequences 9 sequences N1443 30 sequences 378 sequences

b) DHAD distance maps



Run times

(64 GB RAM, 5 threads on 2x 2.6 GHz 14C Xeon VM)











Hybrid ancestors



N51	V L	LF	P	F	L	R	R	R	W	L	L	S	Ρ	Ρ	L	R	R	A	A	G	A	G	R	R	s	A	L
N51_27dLLSPP	V L	LF	P	F	L	R	R	R	W	-	-	-	-	-	L	R	R	A	A	G	A	G	R	R	s	A	L
N2	L L	IF	P	F	L	L	R	R	W	-	-	-	-	-	G	R	R	A	A	G	A	S	R	R	S	A	L
N2_27iLLSPP	L L 16	IF	P	F	L	L	R	R	W	L	L	S	Ρ	Ρ	G	R	R	A	A	G	A	S	R	R	S	A	L 14
	10																										
N5	G L	AI	t v	К	S	Е	L	L	R	L	S	Е	Е	S	G	G	S	G	v	D	L	т	Ρ	L	Ι	S	Ν
N5_153dLSEE	G L	A I	C V	к	S	E	L	L	R	-	-	-	-	s	G	G	s	G	v	D	L	т	Ρ	L	I	S	Ν
N2	E L	ΚF	۰	к	S	Е	М	L	R	-	-	-	-	н	G	G	G	A	F	Ν	Ρ	S	Ρ	I	I	Ν	Ν
N2_152iLSEE	E L	ΚF	۰	К	S	Е	М	L	R	L	S	Е	Е	Н	G	G	G	A	F	Ν	Ρ	S	Ρ	I	I	Ν	Ν
	142/14	43																								17	70/17
N 14																											
N1	L L	sι	. L	Ι	Ρ	Ρ	F	L	L	R	R	W	G	R	R	А	A	G	А	S	R	R	S	Α	L	L	S
N1_19dIPRR	L L	sι	. L	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	-	S	A	L	L	S
	14																										42

GRASP's partial order graphs allow the identification of blocks of content which can be used to create ancestral variants.

CYP2U1 variants shown to fold but with varied substrate selectivity.

c) Activity with luciferin CEE



d) Activity with luciferin ME-EGE



Conclusion

- Ancestral sequence reconstruction is a valuable resource to understand, explore, and utilise evolution
- Large data sets allow us to extend the reach of ASR
- GRASP enables novel experiments on previously unobtainable data set sizes



Acknowledgements



http://grasp.scmb.uq.edu.au

GRASP implementation	CYP2U1 experimental work	DHAD experimental work
Mikael Bodén	Elizabeth Gillam	Volker Sieber
Ariane Mora	Connie Ross	Scott Bottoms
Marnie Lamprecht	Raine Thomson	Jörg Carsten
Julian Zaugg		
Alexandra Essebier	Ross Barnard	GMC experimental work
Brad Balderson	Luke Guddat	Dietmar Haltrich
Rhys Newell	Gary Schenk	Leander Sützl
	Bostjan Kobe	

Burkhard Rost

SCAN ME

Additional slides

Cytochrome P450 2U1 subfamily

Cytochrome P450 enzymes are members of a superfamily of monooxygenases that play a **critical role in metabolism**

CYP2U1 – cytochrome P450 subfamily found across, amphibians, reptiles, mammals, birds, and fish.



CYP2U1 is interesting because –

- No exact established function and substrate specificity
- Previous cytochrome P450 ancestors showed increased stability and promiscuity



Marginal & joint differences



P(n4 = A, n5 = A) = 0.4
P(n4 = A, n5 = C) = 0.3
P(n4 = C, n5 = A) = 0.05
P(n4 = C, n5 = C) = 0.25

Joint reconstruction of node n4 and node n5 Find the highest probability P(n4 = A, n5 = A) = 0.4

Character at n5 is assigned A

Marginal reconstruction of node n5 Sum up all the ways we could get n5=A P(n4 = A, n5 = A) + P(n4= C, n5 = A)= 0.4 + 0.05= 0.45

Sum up all the ways we could get n5=C

P(n4 = A, n5 = C) + P(n4 = C, n5 = C) = 0.3 + 0.25 = 0.55

Character at n5 is assigned C

Marginal & joint differences

1.0

0.8

0.6

0.4

0.2

0.0

Posterior probability distributions from the CYP2U1 CYP2R1 Realigned marginal reconstruction at positions where the marginal and joint reconstructions differ N194 ID: 52 N194 ID: 54 N194 ID: 43 1.0 N194 ID: 64 N194 ID: 17 N194 ID: 26 1.0 1.0 1.0 1.0 0.8 0.8 0.8 0.8 0. 0.6 0.6 0.6 0.6 0.4 0.4 0.4 0.4 0.2 0.2 0.2 0.2 0.0 0.0 0.0 AILMV ADEGHKNPQRST A P Q S T AFGI LNPRSTVWY AEGHKLQRSTVW ADEGI KNPRSTV N194 ID: 181 N194 ID: 175 N194 ID: 73 N194 ID: 78 1.0 N194 ID: 91 N194 ID: 115 1.0 1.0 1.0 1.0 1.0 0.8 0.8 0.8 0.8 0.6 0.6 0.6 0.6 0.6 0.6 0.4 0.4 0.4 0.4 0.4 0. 0 0.2 0.2 0.: 0.2 0.2 0.0 0.0 0.0 ACFHNSTY G S T DEHKNQS V Α L M V AEGKNPQRST N194 ID: 185 N194 ID: 232 N194 ID: 363 N194 ID: 372 N194 ID: 303 1.0 1.0 1.0 N194 ID: 263 1.0 1.0 1.0 0.8 0.8 0.8 0.8 0.8 0. 0.6 0.6 0.6 0.6 0.6 0.4 0.4 0. 0.4 0.4 0.4 0.2 0.2 0.2 0. 0.2 0.2 0.0 0.0 0.0 V 0.0 1 ADEGKMNQRSTV AGILSTV ADEGNST ALPSTV ADEGHKNPQRST N194 ID: 486 N194 ID: 488 N194 ID: 515 N194 ID: 416 1.0 1.0 1.0 0.8 0.8 0.8 0.8 0.6 0.6 0.6 0.6 Amino acid inferred in marginal 0.4 0.4 0.4 0.4 reconstruction 0.2 0.2 0.2 0.2 0.0 0. 00 Amino acid inferred in joint reconstruction V AILPSTV 1 ĸ Q R ACHNRSTY Н

Methods' consensus standard: Generate ancestors that are similar to those of other methods

Data consensus standard: Generate ancestors close to that of the superset



Distance between members and superset ancestor

