Data architecture and workflow for high-throughput genomics

VJ Carey, Channing Lab; C2007 Bioconductor Foundation of N.A.

• views on outreach

- four experimental paradigms
- interface contracts and compliance



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	High-throughput biology,	, epitomized by the ub	iquitous DNA					
Computing preliminarie	es for CDATA-07							
 Students in CDATA 	A-07 are expected to ha	ve at least modest fa	miliarity with R. an or	pen-source system and				
programming lang	juage for data analysis.	We strongly recomm	mend that you get a co	py of R and examine closely				
most if not all of the resources listed below well in advance of your arrival at CSHL.								
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Note that the	re is a system of mailing	a lists to support dia	loque among users an	d developers. Elementary				
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Cold Spring Harbor CDATA-07 [Integrative data analysis for high-throughput biology]

• Strategy – 2 week course (!)

- prominent experimentalists/methodologists lecture most days (e.g., Futcher, Lieb, Wigler, Cheung, Spielman, Baggerly, Irizarry, Carvalho)
- dataset/packages associated with each technique form basis for stat-oriented lectures and labs
- most students are experimentalists (have had NIH and FDA project officers as well); many not very happy with statistics curriculum
- will target advertising at stats and comp-bio departments for 08;

Upstream data structures





Basic outreach approach

- make it easy for practitioners to touch, manipulate, interrogate digital products of exemplary experiments
- big motivation to learn a little R

Integrative container design: Four experimental paradigms

- Useful aim: All relevant information from an experiment or family of related experiments should be contained in a single, aptly named, R variable
- Moderately successful examples:
 - Golub_Merge (but developers still include their own matrix representations in data?)
 - -yeastCC same
 - chr20GGceuRMA (hgfocus + hapmap 700K for Utah CEPH founders)
 - -harbChIP (abuse of ExpressionSet structure; feature-Data includes intergenic sequence data)
 - -neveExCGH (array CGH + u133a on 50 breast cancer cell lines)

Paradigm 1: expression time series

```
> library(yeastCC)
> data(spYCCES)
> spYCCES
ExpressionSet (storageMode: environment)
assayData: 6178 features, 77 samples
 element names: exprs
phenoData
 sampleNames: cln3_40, cln3_30, ..., elu_390 (77 total)
 varLabels and varMetadata description:
    syncmeth: experimental method of synchronization or cyclin induction
   time: in minutes
   phase: Phase of the cell cycle. M: mitosis, G1: gap 1, S: DNA synthesis, G
  gap 2.
featureData
 featureNames: YAL001C, YAL002W, ..., YPR204W (6178 total)
 fvarLabels and fvarMetadata description: none
experimentData: use 'experimentData(object)'
 pubMedIds: 9843569
Annotation: YEAST
```

> experimentData(spYCCES)

Experiment data

Experimenter name: Spellman PT

Laboratory: Department of Genetics, Stanford University Medical Center, Stan Contact information:

Title: Comprehensive identification of cell cycle-regulated genes of the yea URL:

PMIDs: 9843569

Abstract: A 150 word abstract is available. Use 'abstract' method.

> abstract(spYCCES)

[1] "We sought to create a comprehensive catalog of yeast genes whose transcript levels vary periodically within the cell cycle. To this end, we used DNA microarrays and samples from yeast cultures synchronized by three independent methods: alpha factor arrest, elutriation, and arrest of a cdc15 temperature-sensitive mutant. Using periodicity and correlation algorithms, we identified 800 genes that meet an objective minimum criterion for cell cycle regulation. In separate experiments, designed to examine the effects of inducing either the G1 cyclin Cln3p or the B-type cyclin Clb2p, we found that the mRNA levels of more than half of these 800 genes respond to one or both of these cyclins....

Design: sync meth vs sampling times

> table(spYCCES\$sync, spYCCES\$time)[, 1:15] 0 7 10 14 20 21 28 30 35 40 42 49 50 56 60 alpha 1 1 0 1 0 1 1 0 1 0 1 1 0 1 0 cdc15 0 0 1 0 0 0 0 1 0 0 0 0 1 0 0 cdc28 1 0 1 0 1 0 0 1 0 1 0 1 0 1 0 1 clb2 0 0 0 0 0 0 0 0 0 0 2 0 0 0 0 0 cln3 0 0 0 0 0 0 0 1 0 1 0 0 0 0 0 elu 1 0 0 0 0 0 0 1 0 0 0 0 0 0 1

Biology: (declared) phase vs sampling times

>	<pre>table(spYCCES\$phase,</pre>							spl	CCE	ES\$1	time	e)[, 1	:15]	
		0	7	10	14	20	21	28	30	35	40	42	49	50	56	60
	G1	0	0	1	1	1	1	0	0	0	0	0	0	1	0	1
	G2	0	0	0	0	0	0	0	0	0	0	1	0	1	0	0
	М	0	0	0	0	0	0	0	0	0	0	0	1	0	1	1
	M/G1	3	1	1	0	0	0	0	2	0	0	0	0	0	0	0
	S	0	0	0	0	0	0	1	1	1	1	0	0	0	0	0

programming: filtering to see (declared) phase vs sampling times

>	CDC1	5 =	wh	ich	(sp)	YCCI	ES\$a	sync=	=="co	lc15'	')					
>	<pre>table(spYCCES\$phase[CDC15], spYCCES\$time[CDC15])[,</pre>								1:15	5]						
		10	30	50	70	80	90	100	110	120	130	140	150	160	170	180
	G1	0	0	1	1	0	0	0	0	0	0	1	1	1	0	0
	S	0	0	0	0	1	0	0	0	0	0	0	0	0	1	1
	G2	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0
	М	0	0	0	0	0	0	1	1	0	0	0	0	0	0	0
	M/G1	1	1	0	0	0	0	0	0	1	1	0	0	0	0	0

• Exercise: compare samples on a given gene obtained with different synchronization methods with respect to periodicity of expression



PIR1/elu





paradigm 2: ChIP-chip, harbChIP in package:harbChIP

```
ExpressionSet (storageMode: lockedEnvironment)
assayData: 6230 features, 204 samples
  element names: exprs, se.exprs
phenoData
  rowNames: A1 (MATA1), ABF1, ..., ZMS1 (204 total)
  varLabels and varMetadata description:
    txFac: transcription factor symbol from Harbison website CSV fil
  S
featureData
  featureNames: YAL001C, YAL002W, ..., MRH1 (6230 total)
  fvarLabels and fvarMetadata description:
    ID: NA
   PLATE: NA
    •
   REV_SEQ: NA
    (12 \text{ total})
experimentData: use 'experimentData(object)'
  pubMedIds: 15343339
```



qqnorm int.ratio RDS1

Theoretical Quantiles

- Exercise: verify that the motif for RDS1 discovered by Harbison et al, CGGCCG, is present in the most highly bound intergenic regions for that TF.
- Is it present in modestly or weakly bound regions?

>	bigRDS1	=	sort(exprs)	(harbChIP)[,	"RDS1"],	decr =	TRUE)[1:5]
---	---------	---	-------------	--------------	----------	--------	------------

> bigRDS1

YCR107W

YCR107W YCR105W YOR001W YOR258W YGL157W

5.387419 3.743832 2.676375 2.354235 2.317927

> pData(featureData(harbChIP))[names(.Last.value),8:10]

FOR_SEQ REV_SEQ_NO

<NA> <NA>

YGP24747	CCGCTGCTAGGCGCGCCGTGAAAATGCATGTCAAATCTCGGA	YCR105W
YGP34235	CCGCTGCTAGGCGCGCCGTGTAAAAGGTGATTATGTAAAACAAGCG	YOROO1W
YGP34755	CCGCTGCTAGGCGCGCCGTGCAAGCTTTCTCGCATTTCTTT	YOR258W
YGP27527	CCGCTGCTAGGCGCGCCGTGGTATCACGCTAATTGAAGTTTTTTTT	YGL157W
	REV_SEQ	
	<na></na>	YCR107W
	GCAGGGATGCGGCCGCTGACTTCATTTGTTTATCTACCGCTTACATT	YCR105W
	GCAGGGATGCGGCCGCTGACATTTTCTATGCGAAGCCTGATGT	YORO01W
	GCAGGGATGCGGCCGCTGACTTATGATGTTAAAAAGACATGTGTATG	YOR258W

YGL157W GCAGGGATGCGGCCGCTGACTAATTATTTTTGAAACTCTTTTGCAGC

Paradigm 3: genetics of gene expression

```
Samples 58
```

expression assayData
 Storage mode: lockedEnvironment
 featureNames: 1007_s_at, 1053_at, ..., AFFX-r2-P1-cre-3_at, AFFX-r
 Dimensions:
 exprs
Features 8793
Samples 58

> exprs(chr20GGdem)[1:5,1:5]

NA06985 NA06993 NA06994 NA07000 NA07022 1007_s_at 6.236674 5.631134 5.883270 5.791671 5.995744 1053_at 6.535133 6.680420 6.860158 6.298467 6.503476 117_at 4.660155 5.006104 5.018725 4.952051 6.156085 121_at 7.694798 7.331357 7.163441 6.941026 7.361222 1255_g_at 2.831350 2.709704 2.729904 2.723419 3.032477 > snps(chr20GGdem)[1:5,1:5]

NA06985 NA06993 NA06994 NA07000 NA07022

rs4814683	2	0	0	2	1
rs6076506	0	0	0	0	NA
rs6139074	2	0	0	2	1
rs1418258	2	0	0	2	1
rs7274499	0	0	0	0	NA



minor allele count, rs6060535

Genetics of gene expression: analyses, mechanics

- snpScreen is well-documented, can help with focused tests of association between expression and genotypes
- messy, because gene and SNP location data are in various places
- SQLite annotation paradigm will greatly simplify storage and querying of metadata
- other genotype data representations (e.g., Clayton's snpMatrix) will be interfaced

paradigm 4: expression + aCGH

- Neve2006 documents and exemplifies cghExSet class
- ML lab will work with neveCGHmatch; in 2.6 neveExCGH is unified
- logRatios() and exprs(); cloneNames(), cloneMeta()
- NB: the contract for a modeling method segmentation
- Consider rpart as a device for fitting piecewise constant model
- for free: predict on any compliant data frame, x-validated complexity diagnostic



sample 3 ; BaA

sample 4 ; Lu



kB on chr 17

log ratio

sample 5 ; Lu



sample 6 ; BaB





Integrative container behavior

- Useful idiom: X[G, S] is a selection
 - -G is a reporter selection predicate
 - -S is a sample selection predicate
- can we entertain:
 - X[clonesNear("CPNE1"), homRare("rs6060535")]

and would we want to? This would select copy number data near a certain gene for samples that have a certain genotype. The string parameters might work, or we might need hugo("CPNE1"), e.g., to specify semantics

 awaited maturity of SQLite annotation, now can experiment

contracts of statistical modeling procedures

- formula interface is useful
- predict(), plot(), coef(), residuals(), should exist and make sense (comply with reasonable expectations)
- many things that are fitting models (e.g., normalization functions) do not attempt this
- self-discipline is hard; software helpers (e.g., stub generators) are in wide use in other languages

transparency and agnosticism

• DR Cox, AAS 2007

It is interesting and perhaps surprising that J. W. Tukey, who had an extraordinarily wide-ranging knowledge of the natural sciences down to fine detail, favored largely ignoring that knowledge in the main phases of analysis, introducing it only in the final stages of interpretation.

transparency and agnosticism

• JD Watson, The Double Helix, ch 28.

Maurice, in a lab devoid of structural chemists, did not have anyone to tell him that all the textbook pictures were wrong.

conclusions

- main jobs: data capture, removal of irrelevant noise, interrogation/modeling, reporting
- multiassay containers coordinate and provide access to diverse measurements
- concise interrogators: generic methods that capitalize on simultaneous access to assay, phenotype, and biological metadata
- modeling functions should respect a well-established tradition of input and return capabilities
- tremendous development of container infrastructure: S. Falcon, M. Morgan, others