Multiple sequence alignment accuracy estimation and its role in creating an automated bioinformatician

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Tunable parameters



Tunable parameters

Quant

=========

Perform dual-phase, mapping-based estimation of transcript abundance from RNA-seq reads

salmon quant options:

basic options:

-v [version]	print version string
-h [heln]	produce help message
_i [index] arg	Salmon index
-] [libType] arg	Format string describing the library type
-r [unmatedBeads] arg	List of files containing unmated reads of (e.g. single-end reads)
-1 [mates1] arg	File containing the #1 mates
-2 [mates2] arg	File containing the #2 mates
-o [output] arg	Output quantification file.
discardOrphansQuasi	[Quasi-mapping mode only] : Discard orphan mappings in quasi-mapping mode. If this flag is passed then only paired mappings
arboar aor phano daabr	will be considered toward quantification estimates. The default behavior is to consider orphan mappings if no valid paired
	mappings exist. This flag is independent of the option to write the orphaned mappings to file (writeOrphanLinks).
allowOrphansFMD	[FMD-mapping mode only] : Consider orphaned reads as valid hits when performing lightweight-alignment. This option will
	increase sensitivity (allow more reads to map and more transcripts to be detected), but may decrease specificity as orphaned
	alignments are more likely to be spurious.
seqBias	Perform sequence-specific bias correction.
gcBias	[beta for single-end reads] Perform fragment GC bias correction
-p [threads] arg	The number of threads to use concurrently.
incompatPrior arg	This option sets the prior probability that an alignment that disagrees with the specified library type (libType) results
	from the true fragment origin. Setting this to 0 specifies that alignments that disagree with the library type should be
	"impossible", while setting it to 1 says that alignments that disagree with the library type are no less likely than those
	that do
-g [geneMap] arg	File containing a mapping of transcripts to genes. If this file is provided Salmon will output both quant.sf and
	quant.genes.sf files, where the latter contains aggregated gene-level abundance estimates. The transcript to gene mapping
	should be provided as either a GTF file, or a in a simple tab-delimited format where each line contains the name of a
	transcript and the gene to which it belongs separated by a tab. The extension of the file is used to determine how the file
	should be parsed. Files ending in '.gtf', '.gff' or '.gff3' are assumed to be in GTF format; files with any other extension
	are assumed to be in the simple format. In GTF / GFF format, the "transcript_id" is assumed to contain the transcript
	identifier and the "gene_id" is assumed to contain the corresponding gene identifier.
<pre>-z [writeMappings] [=arg(=-)]</pre>	If this option is provided, then the quasi-mapping results will be written out in SAM-compatible format. By default, output
	will be directed to stdout, but an alternative file name can be provided instead.
meta	If you're using Salmon on a metagenomic dataset, consider setting this flag to disable parts of the abundance estimation model
	that make less sense for metagenomic data.
advanced options:	
alternativeInitMode	[Experimental]: Use an alternative strategy (rather than simple interpolation between) the online and uniform abundance
	estimates to initalize the EM / VBEM algorithm.

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Motivation

Almost all pieces of scientific software have tunable parameters.

- Their settings can greatly impact the quality of output.
- Default parameters are best on average but may be bad in general.

default	<pre> yl-lhqflspssnqrtdqyggsvenrarlvlevvdavcnewsad-RIGIRVSPigtfq kP-LGVKLPPyfdlvhfdimaeilnqfpltyvsnv-nsignglfidpeaesv yl-lnqfldphsntrtdeyggsienrarftlevvdalveaighe-KVGLRLSPygvfn yl-plqflnpyynkrtdkyggslenrarfwletlekvkhavgsdcAIATRFGVdt kvPLYVKLSPnv-tdivpiakaveaagadgltmintlmgvrfdlktrqp</pre>	· ·
alternate	<pre> gsvenrarlvlevvdavcnewsad-RIGIRVSPigtfqnvdngpneeadalyl ydfeatekllkevftfftk-PLGVKLPPyfdlvhfdim . gsienrarftlevvdalveaighe-KVGLRLSPygvfnsmsggaetgivaqyayvage . gslenrarfwletlekvkhavgsdcAIATRFGVdtvygpgq . tdpevaaalvkackavskv-PLYVKLSPnvtdivpiaka .</pre>	• • • • • •

Motivation

Almost all pieces of scientific software have tunable parameters.

- Their settings can greatly impact the quality of output.
- Default parameters are best on average but may be bad in general.
- The most interesting problems are rarely "average".
- Mis-configuration can lead to missed or incorrect conclusions.

Can we find a parameter choice that is best for a given input?

Parameter advising

Given an input an advisor

- uses an advisor set of parameter choice vectors to obtain candidates, then
- returns the most accurate of these solutions based on the accuracy estimation.



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Transcript assembly advising

Reference is available, therefore Area Under the ROC Curve was used as the accuracy estimator



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How can we use Parameter Advising when a reference isn't available?

Multiple sequence alignment

Given

- a set of sequences $\{S_1, S_2, \dots, S_m\}$, and
- an alignment objective function
- find an $m \ge n$ matrix
 - where each row represents one sequence from the set with inserted gaps, and
 - is optimal under the objective function.



Motivation

Multiple sequence alignment is a fundamental problem in bioinformatics.

- multiple sequence alignment is NP-Complete
- many popular aligners for multiple sequence alignment
- each aligner has many parameters whose values affect the alignment that is output

Alignment accuracy is measured with respect to a reference alignment.

reference alignment	computed alignment
··· a D E h s ···	a <mark>D E</mark> h – s
··· d <mark>S R –</mark> d ···	··· d <mark>S R – –</mark> d ···
··· a <mark>SH</mark> lt ···	a <mark>S-H</mark> lt

 accuracy is the fraction of substitutions from the reference that are in the computed alignment,

Alignment accuracy is measured with respect to a reference alignment.



- accuracy is the fraction of substitutions from the reference that are in the computed alignment,
- measured on the core columns of the reference.

Our estimator Facet ("Feature-based ACcuracy EsTimator")

- estimates accuracy by a polynomial on feature functions,
- efficiently learns the polynomial coefficients from examples,
- uses novel features that are efficient to evaluate.

The estimator E(A) is a polynomial in the feature functions $f_i(A)$.

linear estimator

$$E(A) := \sum_{i} c_i f_i(A)$$

quadratic estimator

$$E(A) := \sum_{i} c_{i} f_{i}(A) + \sum_{i} \sum_{j} c_{ij} f_{i}(A) f_{j}(A)$$

Learning the estimator

We learn the estimator using examples consisting of

- an alignment, and
- its associated true accuracy.

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Learning finds optimal coefficients that either fit

- accuracy values of the examples, or
- accuracy differences on pairs of examples,
- by solving a linear or quadratic program.

We use protein alignment feature functions that

- are fast to evaluate,
- measure novel properties,
- use non-local information,
- involve secondary structure.

Features based only on the input alignment

- Amino Acid Identity
- Average Substitution Score
- Information Content
- ...

There are three types of secondary structure

- a-helix,
- β-strand,
- coil.



http://www.ebi.ac.uk/training/online/

Features using predicted secondary structure

- Secondary Structure Percent Identity
- Secondary Structure Agreement
- Secondary Structure Blockiness
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A block ${\cal B}$ in alignment ${\cal A}$ is

- an interval of at least l columns,
- a subset of at least k rows,
- with the same secondary structure for all residues in B.



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Theorem (Evaluating Blockiness)

Blockiness can be computed in O(mn) time, for an alignment with *m* rows and *n* columns.



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Algorithm



S F E D Ε F Τ S E G Ε Ε E Τ V Α Μ Ι G Η t L Ν Ρ Ι Τ L E t G Ι Ν L Η Ρ Η Τ L

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Algorithm



minimum width l = 3

—	Ε	F	Е	D	S	Е	F	т	S
-	G	A	Е	Е	М	Е	Т	V	I
t	G	Ν	Н	Р	I	Е	Т	L	L
t	G	Ν	Н	Р	I	H	т	L	L

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t	G	Ν	Η	Р	I	Е	т	L	L
t	G	N	Н	Р	I	H	т	L	L

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_	Ε	F	Е	D	S	Е	F	Т	S
_	G	А	Е	Е	М	Е	Т	V	I
t	G	N	Η	Ρ	I	Е	т	L	L
t	G	N	Н	Р	I	Н	т	L	L

Theorem (Evaluating Blockiness)

Blockiness can be computed in O(mn) time, for an alignment with *m* rows and *n* columns.



- Graph construction takes O(mn) time.
- Graph has O(n) nodes, O(ln) edges
- Longest path takes O(n) time.

Best features trend well with accuracy.



Facet estimator has less spread than its features.

For parameter advising, an estimator should have high slope and low spread.



Facet's slope and spread is best for advising

Parameter advising

A parameter advisor has two components:

- an accuracy estimator, and
- a set of candidate parameter choices.



Given accuracy estimator, what is the *optimal set* of parameter choices?

Advisor Set problem

For the Advisor Set problem the input is

- cardinality bound k,
- universe of parameters choices U,
- a collection of examples, each with a
 - true accuracy,
 - estimator value, and
 - benchmark weight.

Advisor Set problem

The output is

an optimal set P ⊆ U of parameter choices
with |P| ≤ k, that maximizes the objective function

$$\sum_{\text{benchmarks } i} w_i \operatorname{Accuracy}_i(P)$$

• where $Accuracy_i(P)$ is the true accuracy for benchmark *i* of the example chosen by an advisor that uses advisor set *P*

Advisor Set problem

THEOREM (Problem Complexity)

The Advisor Set problem is NP-complete.

- Reduction is from the Dominating Set problem
- Polynomial-time solvable for fixed \boldsymbol{k}
- Optimal oracle sets can be found for all k in practice by integer linear programming

Approximation algorithm

THEOREM (Approximation Algorithm)

A greedy approach yields an $\frac{\ell}{k}$ -approximation algorithm for Advisor Set, for constant $\ell < k$.

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The approximation ratio $\frac{\ell}{k}$ is tight.

Experimental results

To evaluate the accuracy of advising, we consider:

- Facet and other estimators,
- over 800 benchmarks,
- evaluated with k-fold cross validation.

Experimental results

Advising performance for various estimators



For parameter advising Facet has the highest accuracy

Experimental results

Average accuracy of advisors by difficulty bin



Boosts the accuracy on the hardest bins by almost 20%

Summary

- Parameter Advising can be used as a framework used to find input-specific parameter choices.
- Facet has the best performance for advising multiple sequence alignment.
- The Facet framework can be used to learn estimators for domains where features can be extracted.

References

Multiple Sequence Alignment Advising

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- Kececioglu and DeBlasio. Accuracy estimation and parameter advising for protein multiple sequence alignment. Journal of Computational Biology, 20(4), April 2013.

Transcript Assembly

 DeBlasio and Kingsford. Automatically eliminating errors induced by incorrect parameter choices for transcriptome assembly. *bioRxiv* 2018(342865). *under review.*

<u>Acknowledgments</u>

Software

facet.cs.arizona.edu

github.com/Kingsford-Group/scallopadvising

Contact

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Collaborators

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