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#### November 1, 2022

#### Overview.

LACE 2.0 includes: (a) an easy-to-use module for data processing and organization, which allows one to test and adjust several quality metrics and relevance filters, and to annotate gene variants; (b) a module for the interactive visualization of the inferred cancer evolution models, which returns both the *longitudinal clonal tree* and the *fishplot*, and allows one to examine the different parts of the model (i.e., clones and temporal relations), also by querying external databases such as Ensembl.org. Importantly, LACE 2.0 requires a minimum set of parameters, which are sufficient to mitigate the aforementioned noise sources and infer reliable models of cancer evolution.

In this vignette, we give an overview of the Shiny App by presenting its main functions.

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LACE 2.0 is a new release of the LACE R package. LACE 2.0 is capable of performing clonal evolution analyses for single-cell sequencing data including longitudinal experiments. LACE 2.0 allows to annotate variants and retrieve the relevant mutations interactively based on user-defined filtering criteria; it infers the maximum likelihood clonal tree, cell matrix attachments and false positive/negative rates using boolean matrix factorization. Furthermore, LACE 2.0 allows to investigate cancer clonal evolution under different experimental conditions and the occurrence of single mutations which can be queried via *ensembl* database.

# 1 Installation of LACE 2.0 R package

The package is available on GitHub and Bioconductor. LACE 2.0 requires  $\mathsf{R}>4.1.0$  and Bioconductor. To install Bioconductor run:

if (!require("BiocManager", quietly = TRUE))

install.packages("BiocManager")

To install LACE 2.0 run:

remotes::install\_github("https://github.com/BIMIB-DISCo/LACE", dependencies = TRUE)

LACE 2.0 uses *Annovar* and *Samtools suite* as back-ends for variant calling annotation and depth computation, respectively. Please refer to the next section to install them.

## 2 Installation of other required softwares

Annovar is a widely used variant calling software freely available upon registration to their website at https://annovar.openbioinformatics.org/en/latest/. The package contains *Perl* scripts and variant calling annotation reference databases for the human species. For other databases, please refer to their website. If the scripts are installed in binary search path, then LACE 2.0 will detect them automatically.

Perl (https://www.perl.org/) is required to run Annovar.

Samtools suite is a standard set of tools and libraries to handle SAM/BAM/BED file format and perform a variety of common operations on sequencing data. It is freely available at http://www.htslib.org/ and https://github.com/samtools/htslib. To install Samtools follow the instructions in their website.

# 3 Running LACE 2.0

To start LACE 2.0 user interface run: library(LACE) LACEview()

# 4 Using LACE 2.0

LACE 2.0 has been thought to be used on single cell sequencing data for which it is available variant calling data in standard VCF format and binary aligned data in standard BAM format.

The user is provided with an interface to initiate a project and to set filter thresholds after which annotation of variants, filtering of data and depth at variant sites are retrieved. Both annotation and depth derivation are computationally expensive steps. LACE 2.0 reduces possible re-computation by detecting parameter variations of the user interface and by comparison of the timestamps of interface state, inputs and outputs. Intermediary and final data are stored in the designated folders.

The operation is followed by the possibility for the user to select variants which are drivers of the understudied biological problem.

At this point, the inferential step is executed so that the most likelihood longitudinal clonal tree and clonal prevalences are retrieved together with the best set of false positive/negative rates among those provided by the user.

The results are displayed via an interactive interface.

## 5 Interface

The interface is divided in two parts interleaved by variant selection and inferential computation parts: the processing interface and the results interface.

### 5.1 Project creation

To begin the clonal analyses the user needs to create a project by choosing a folder and a meaningful name for the project.

If the project path does not contain former LACE projects, it asks the user to create a new one ("Create project") and all intermediate and final results will be stored in structured sub-folders of the chosen project path.

LACE 2.0 automatically recognizes if the selected path is a LACE project and proposes the user to load it ( Load project )

### 5.2 Sidebar and Demos

The sidebar is divided in Projects and Current Project. Projects contain a set of recently opened projects which can be reloaded by selecting the respective project name.

	LACE 2.0									
🗞 P	Projects	•								
"ତ ।	Recent Projects	<								
8	Demo Projects	<								
>	<ul> <li>Melanoma dataset (Ra</li> </ul>	mbow								
>	Small dataset									
<b>@</b> c	Current Project	<								

Under Projects, there is the sub-menu Demos. LACE 2.0 provides the user with two demos:

- Melanoma dataset [Rambow, Florian, et al. "Toward minimal residual disease-directed therapy in melanoma." Cell 174.4 (2018): 843-855.] containing single cell variant annotations and depths for a total of 674 cells sampled at 4 different time points:
  - 1. sample before treatment
    - BRAF inhibitor treatment
  - 2. sample after 4 days
  - 3. sample after 28 days
  - 4. sample after 57 days
- Small dataset containing only 3 cells per 4 sampling time points.

Current Project allows the user to save and load the parameters of the whole project or of the active tab.

### 5.3 Processing interface

#### 5.3.1 Single Cell Metadata

In order to perform clonal analyses it is required to provide some information about the experiment such as the cell IDs which are used to retrieve the VCF/BAM files and the sample name each cell belongs to. These metadata are generally included with the experimental

sequencing data as part of the library preparation. LACE 2.0 accepts tsv/csv tabular formats with headers and rds standard data R format. Metadata file can include more information relative to the experimental setup and the platform used.

Sin	gle cell r	netada	ta info		
belon	ig to is require ells used in the	d. The iden	ifiers are used	to identif	iment. At a minimum, information about the cells' identifiers and the sample they y the file names to include in the anlysis and the samplings represent the time points section between the identifiers present in the metadata and the BAM files stored in th
You n	nay have to cre	ate such a	ile from scratc	h. See the	(i) button below.
lf you	have loaded o	ne of the 'o	lemo projects',	changes (	on this tab are not considered.
				•	
Meta	data file		ata_info.rds	•	
	/Rambow	_dataset/d	ata_inio.rds		
C	d column			•	
	d column			0	
<b>Cell i</b> Rur				•	
				•	
	n	2		•	
Rur	n Run			•	
Rur 1	<b>Run</b> SRR7424152	3		•	
Rur 1 2	Run SRR7424152 SRR7424153	- }		•	
Rur 1 2 3	Run           SRR7424152           SRR7424152           SRR7424152	5		•	
Rur 1 2 3 4	Run SRR7424152 SRR7424153 SRR7424154 SRR7424154	- } ;		•	
Rur 1 2 3 4 5	Run           SRR7424152           SRR7424153           SRR7424154           SRR7424154           SRR7424154           SRR7424154           SRR7424154	- } ;		•	
Run 1 2 3 4 5 6	Run           SRR7424152           SRR7424153           SRR7424154           SRR7424154           SRR7424154           SRR7424154           SRR7424154	- } ;		•	
Run 1 2 3 4 5 6	Run SRR7424152 SRR7424153 SRR7424154 SRR7424154 SRR7424155 SRR7424155 SRR7424155	- } ;		•	

The user should identify and select the ID and sampling columns. The user can also reorder the sampling names

Time column	0	
Age	•	Sampling time points
	•	
Sampling points Drag the time points in chronological orde (before the smaller times)	r	Select and drug one or more sampling points to order them chronologically.
before treatment	•	
4d on treatment		
28d on treatment		
57d on treatment	-	

so to recreate the chronological order of longitudinal events of the experiment, or to explore different orders.

#### 5.3.2 Annotations

Annotations of variant calling data is performed using *Annovar* as back-end. Each mutation of each cell is annotated based on the annotation database used. *Annovar* provides a database for the human species useful to tag cancer mutations and whenever possible their functional effect.

If *Annovar*'s *Perl* scripts are available in the standard binary paths, they are detected and the *Annovar* folder field is autocompleted; otherwise, the user should provide the folder containing the *Perl* scripts. The user should also provide the database to use for the annotation and the folder containing the variant calling files (VCF).

#### 5.3.3 Variant filtering

All types of single cell sequencing data is characterized by various sources of noise which depends on the technology used. Detected mutation might be characterized by low quality score or low statistical power. Some mutations can be neglected, while others might cause relevant effects, especially on exonic regions where variations can result in changes of the translational process and modify their functional form.

In order to avoid small and possibly spurious fluctuations on sequencing data, variants can be filtered if they have low supporting evidences. The user can set the minimum values for:

- 1. the number of reads supporting the alternative alleles in a cell,
- 2. the frequency of the minor allele for each referenced SNP included in a default global population,
- 3. the cell frequency per sample showing the mutation at the same site.

you have loaded one of the 'demo pro		ext steps.	se the alternative frequency, min functional exonic SNVs to conside	ninimum
	ojects', changes on thi	s tab are no	ot considered.	
ternate frequency	0			
2				
AF	0			
0.01				
ariant frequency	-			
0.05 Konic variant function annotations	6			
	6		Neglected exonic variants	
<b>konic variant function annotations</b> Choose the variant functions for infe	6	A	Neglected exonic variants	
<b>conic variant function annotations</b> Choose the variant functions for infe Considered exonic variants	6	Î		
choose the variant function annotations Choose the variant functions for infe Considered exonic variants stopgain	6	Î	synonymous SNV	
Choose the variant function annotations Choose the variant functions for infe Considered exonic variants stopgain stoploss	i erential analysis		synonymous SNV	
Choose the variant function annotations Choose the variant functions for infer Considered exonic variants stopgain stoploss nonframeshift insertion	i erential analysis		synonymous SNV	

(+)	Explanation
frameshift insertion frameshift deletion frameshift block substitution	an insertion of one or more nucleotides that cause frameshift changes in protein coding sequence a deletion of one or more nucleotides that cause frameshift changes in protein coding sequence a block substitution of one or more nucleotides that cause frameshift changes in protein coding sequence

Annotation	
(+)	Explanation
stopgain	a nonsynonymous SNV, frameshift insertion/deletion, nonframeshift insertion/deletion or block substitution that lead to the immediate creation of stop codon at the variant site. For frameshift mutations, the creation of stop codon downstream of the variant will not be counted as "stopgain"!
stoploss	a nonsynonymous SNV, frameshift insertion/deletion, nonframeshift insertion/deletion or block substitution that lead to the immediate elimination of stop codon at the variant site
nonframeshift insertion	an insertion of 3 or multiples of 3 nucleotides that do not cause frameshift changes in protein coding sequence
nonframeshift deletion	a deletion of 3 or mutliples of 3 nucleotides that do not cause frameshift changes in protein coding sequence
nonframeshift block substitution	a block substitution of one or more nucleotides that do not cause frameshift changes in protein coding sequence
nonsynonymous SNV	a single nucleotide change that cause an amino acid change
synonymous SNV	a single nucleotide change that does not cause an amino acid change
unknown	unknown function (due to various errors in the gene structure definition in the database file)

#### (+) see Annovar

### 5.3.4 Single cell sampling depth

The number of reads per SNV site represents an optimal filter to retrieve relevant mutations. Depth at specific sites is usually not provided in standard alignment or variant calling pipelines and it is computationally expensive to retrieve. The user should provide the folder with aligned data and, if not found already, the *samtools* executable folder location to compute depth only on sites passing the variant filters set in the previous tab.

LACE 2.0		Project: Rambow_dataset
	SC sampling depths	
Single cell sampling dept	h	
	nes and are computationally expensive. Pr	tions. Depth at specific sites is usually not provided in ovide the folder with aligned data and, if not found filters set in the previous tab.
If you have loaded one of the 'demo project	s', changes on this tab are not considered.	
Samtools executable folder ///usr/local/bin	0	
BAMs folder /Rambow_dataset/bam	0	

### 5.4 Selection of relevant variants

Not all mutations are distinctive of the disease or experiment understudied. Identifying relevant and driver variants allows to reproduce a more significant longitudinal clonal tree. The user can select a set of filters based on gold standards and other analyses such as:

- 1. the minimum number of reads at given mutation site
- 2. the maximum number of missing data per gene
- 3. the minimum median depth per locus
- 4. the minimum median depth supporting the mutation
- 5. subset of known genes

	Variants
/ariant filters	
Relevant variants allows to reproduce more variants as part of the processing step.	significant longitudinal clonal tree for the experiment. Apply some filters to retrieve relevant
	ssing right now, and then, interactively modify filters and see the results on selected variants as t id go next to perform processing and inference all together.
Ainimum depth	θ
8	
Aax missing value	Θ
0.3	
ite minumum median depth:	0
8	
Autation minimum median deapth	0
5	
	0
Select variant genes	
Select variant genes ARPC2 × CCT8 × PRAME × COL1A2 CYCS × HNRNPC × PCBP1 × RPL5	2 ×

Finding the parameters to select variants is not an easy task, and the user might not know in advance how to choose the best set of filters. Hence, the user can press "Select variants" ( select variants ) to perform the aforementioned computations on VCF and BAM files to derive all the necessary aggregated information on the sampled cells. Afterward, the user is presented with a interactive live preview of variants passing the filters (including relevant parameters which can help in the selection) while values are changed.

0.3				
ite minu	mum media	n depth:		
8				
lutation	minimum n	andian doan	*h	
	minimum n	nedian deap	th	
lutation	minimum n	nedian deap	th	
	minimum n	nedian deap	th	
5		nedian deap	th	
5 elect var	iant genes			
5 elect var	iant genes	PRAME ×		к

Show	Show 10 * entries Search:													
	Gene 💧	Chr 🕴	PosStart ≬	PosEnd ≬	REF ≬	ALT 🕴	FreqT1 🕴	FreqT2 ≬	FreqT3 🍦	FreqT4 ≬	MedianDepth 🔶	MedianDepthMut 🍦	ResultantMeanDepth 🔅	ResultantVarDepth 🔅
1	ARPC2	2	218249894	218249894	С	т	0.788	0.67	0.79	0.75	12	6	11.266	65.155
2	CCT8	21	29063389	29063389	G	А	0.315	0.275	0.395	0.281	14	5	8.774	52.891
3	CCT8	21	29066994	29066994	-	т	0.027	0.037	0.062	0.016	15	6	9.932	80.682
4	COL1A2	7	94422978	94422978	С	А	0.493	0.394	0.543	0.234	14	6	9.395	113.918
5	CYCS	7	25123715	25123715	-	TT	0.007				12	5	7.408	39.983
6	CYCS	7	25123786	25123786	с	т	0.699	0.606	0.667	0.641	9	5	6.728	35.084
7	HNRNPC	14	21211836	21211836	-	т	0.041	0.046	0.062	0.039	16	5	11.545	81.844
8	PRAME	22	22551005	22551005	т	А	0.541	0.606	0.704	0.484	16	6	13.004	106.337
9	RPL5	1	92837514	92837514	с	G	0.253	0.009	0.049	0.281	31.5	5	29.902	203.518
Show	ing 1 to 9 of 9	9 entries											Prev	rious 1 Next

Run Select variants

#### 5.5 Inference

The inferential tab allows the user to set all the parameters to solve the boolean matrix factorization problem and estimate the model parameters of the Bayesian model using a MCMC to maximize the likelihood.

Inferential step uses the following set of parameters:

- 1. Learning rate
- 2. False positive rates for each sample
- 3. False negative rates for each sample
- 4. Number of iterations in each MCMC search
- 5. Number of restart for the MCMC 5 Early stopping number of iterations with no growing likelihood
- 6. Number of parallel processes
- 7. Random seed to recreate simulations
- 8. Initialize the clonal tree randomly
- 9. Marginalize the cell attachment matrix
- 10. Keep equivalent solutions and return all of them
- 11. Check indistinguishable event and remove them
- 12. Estimate error rates of MCMC moves
- 13. Show results

Number of iterations	0
10000	
Number of restarts	6
50	
Early stopping	6
500	
Number of parallel processes	0
10	
Seed	6
1121	
Random tree intialization	0
Marginalize	0
Keep equivalent	0
Check indistinguishable	0
Error move	0
Show results in interface	6

#### Run

Run LACE

The user must insert at least one false positive rate and one false negative rate value for each sample. By double clicking on any cell belonging to the row "add row", the user can insert a new set of false or positive rates for each sample. During the inferential step, the maximization of the likelihood for each set of rates is performed. The best results are returned.

			Inference	e
Longitudinal tree inference parameters				
egative 'beta' rate fo	or each sampling time point. M	ultiple sets of alpha and beta	ial step. Add, at least, one false can be added by inserting value part of the inference and return	s on the 'add row' cells. If
	ave your current configuration, e variant and depth computati		er the longitudinal clonal tree. If Is previously not included.	the metadata, and filters d
earning rate	0			
1				
sert a set of false po			row to add further sets of rates	to be used in the inference.
nsert a set of false po	ositive rates, one $lpha$ for each sar		row to add further sets of rates 28d on treatment \$	to be used in the inference. 57d on treatment 🖗
nsert a set of false po	ositive rates, one $lpha$ for each sar ntries	npling time point. Fill the last		
nsert a set of false po	ositive rates, one $\alpha$ for each san tries <b>before treatment</b>	mpling time point. Fill the last	28d on treatment 🍦	57d on treatment 🍦
nsert a set of false po showing 1 to 3 of 3 en alpha_1	ositive rates, one α for each san ntries <b>before treatment</b>	npling time point. Fill the last 4d on treatment 0.01	<b>28d on treatment</b> $\Rightarrow$ 0.01	<b>57d on treatment</b> 0.01
nsert a set of false po showing 1 to 3 of 3 en alpha_1 alpha_2 add row	ositive rates, one α for each san htries <b>before treatment ♦</b> 0.02 0.10	npling time point. Fill the last 4d on treatment 0.01 0.05	<b>28d on treatment</b> $\Rightarrow$ 0.01	<b>57d on treatment</b> 0.01
alpha_1 alpha_2 add row	sitive rates, one α for each sar ntries before treatment ♦ 0.02 0.10	Ad on treatment +	<b>28d on treatment</b> $\Rightarrow$ 0.01	57d on treatment ♦ 0.01 0.05
alpha_1 alpha_2 add row	esitive rates, one α for each sar ntries before treatment 0.02 0.10 0.10 0.2 0.10	Ad on treatment +	28d on treatment 🖗 0.01 0.05	57d on treatment ♦ 0.01 0.05
nsert a set of false po showing 1 to 3 of 3 en alpha_1 alpha_2 add row False negative rates nsert a set of false ne	esitive rates, one α for each sar ntries before treatment 0.02 0.10 0.10 0.2 0.10	Ad on treatment +	28d on treatment 🖗 0.01 0.05	57d on treatment ♦ 0.01 0.05
nsert a set of false po howing 1 to 3 of 3 en alpha_1 alpha_2 add row alse negative rates nsert a set of false ne	esitive rates, one α for each sar htries before treatment 0.02 0.10 egative rates, one β for each sar htries	mpling time point. Fill the last 4d on treatment 0.01 0.05 mpling time point. Fill the last	28d on treatment 🖗 0.01 0.05	57d on treatment 0.01 0.05 to be used in the inference.
nsert a set of false po showing 1 to 3 of 3 en alpha_1 alpha_2 add row False negative rates nsert a set of false ne showing 1 to 3 of 3 en	esitive rates, one α for each sar htries before treatment 0.02 0.10 egative rates, one β for each san htries before treatment \$	mpling time point. Fill the last 4d on treatment 0.01 0.05 mpling time point. Fill the last 4d on treatment 4d on treatment	28d on treatment 🖗	57d on treatment 0.01 0.05

Press "Run LACE" to perform all computation and estimation steps and visualize the results.

### 5.6 Longitudinal display and outputs interface

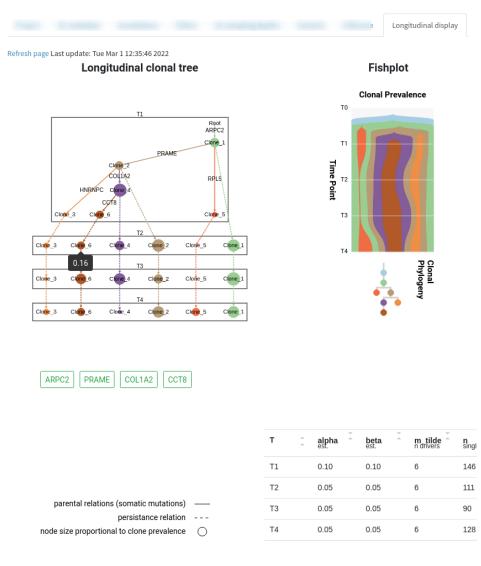
The longitudinal display tab shows the longitudinal clonal tree, the fishplot and the clonal tree, together with the best false positive and negative rate parameters.

The longitudinal clonal tree is placed on the top left side. Continuous edges are used to represent parental relations between clones, while dashed edges show persistence relations. The size of the nodes is proportional to the clonal prevalence. Moving the mouse over a node, it is possible to see the clonal prevalence value. While moving the mouse over the mutation labels, it shows the complete set of mutations for that clone under the longitudinal clonal tree.

The fishplot is on the top right side of the tab. Passing the mouse over the ribbons, the clonal prevalences of the clones at different time points are displayed. Similarly, with the mouse over the time tags, the clonal prevalences of all clones at the specified time point are visualized.

By clicking on a mutation label of the longitudinal clonal tree as well as on a fishplot ribbon or a clonal tree node, the details of the mutation characterizing the clone can be queried and retrieved from the *Ensembl* database.

The bottom part of the tab is used to display the tree legend and the best estimated set false positive/negative rates.



### 6 sessionInfo()

- R version 4.2.1 (2022-06-23), x86\_64-pc-linux-gnu
- Locale: LC\_CTYPE=en\_US.UTF-8, LC\_NUMERIC=C, LC\_TIME=en\_GB, LC\_COLLATE=C, LC\_MONETARY=en\_US.UTF-8, LC\_MESSAGES=en\_US.UTF-8, LC\_PAPER=en\_US.UTF-8, LC\_NAME=C, LC\_ADDRESS=C, LC\_TELEPHONE=C, LC\_MEASUREMENT=en\_US.UTF-8, LC\_IDENTIFICATION=C
- Running under: Ubuntu 20.04.5 LTS

- Matrix products: default
- BLAS: /home/biocbuild/bbs-3.16-bioc/R/lib/libRblas.so
- LAPACK: /home/biocbuild/bbs-3.16-bioc/R/lib/libRlapack.so
- Base packages: base, datasets, grDevices, graphics, methods, stats, utils
- Other packages: LACE 2.2.0, SummarizedExperiment 1.28.0, knitr 1.40
- Loaded via a namespace (and not attached): AnnotationDbi 1.60.0, Biobase 2.58.0, BiocFileCache 2.6.0, BiocGenerics 0.44.0, BiocManager 1.30.19, BiocStyle 2.26.0, Biostrings 2.66.0, DBI 1.1.3, DT 0.26, DelayedArray 0.24.0, GenomeInfoDb 1.34.0, GenomeInfoDbData 1.2.9, GenomicRanges 1.50.0, IRanges 2.32.0, KEGGREST 1.38.0, Matrix 1.5-1, MatrixGenerics 1.10.0, R6 2.5.1, RColorBrewer 1.1-3, RCurl 1.98-1.9, RSQLite 2.2.18, Rcpp 1.0.9, RcppTOML 0.1.7, RcppZiggurat 0.1.6, Rfast 2.0.6, S4Vectors 0.36.0, XML 3.99-0.12, XVector 0.38.0, assertthat 0.2.1, biomaRt 2.54.0, bit 4.0.4, bit64 4.0.5, bitops 1.0-7, blob 1.2.3, bsplus 0.1.3, cachem 1.0.6, callr 3.7.2, cli 3.4.1, codetools 0.2-18, compiler 4.2.1, configr 0.3.5, crayon 1.5.2, curl 4.3.3, data.table 1.14.4, data.tree 1.0.0, dbplyr 2.2.1, digest 0.6.30, doParallel 1.0.17, dplyr 1.0.10, ellipsis 0.3.2, evaluate 0.17, fansi 1.0.3, fastmap 1.1.0, filelock 1.0.2, foreach 1.5.2, fs 1.5.2, generics 0.1.3, glue 1.6.2, grid 4.2.1, highr 0.9, hms 1.1.2, htmltools 0.5.3, htmlwidgets 1.5.4, httpuv 1.6.6, httr 1.4.4, igraph 1.3.5, ini 0.3.1, iterators 1.0.14, isonlite 1.8.3, later 1.3.0, lattice 0.20-45, learnr 0.11.1, lifecycle 1.0.3, logr 1.3.3, lubridate 1.8.0, magrittr 2.0.3, markdown 1.3, matrixStats 0.62.0, memoise 2.0.1, mime 0.12, parallel 4.2.1, pillar 1.8.1, pkgconfig 2.0.3, png 0.1-7, prettyunits 1.1.1, processx 3.8.0, progress 1.2.2, promises 1.2.0.1, ps 1.7.2, purrr 0.3.5, rappdirs 0.3.3, readr 2.1.3, rlang 1.0.6, rmarkdown 2.17, rprojroot 2.0.3, shiny 1.7.3, shinyBS 0.61.1, shinyFiles 0.9.3, shinydashboard 0.7.2, shinyjs 2.1.0, shinythemes 1.2.0, shinyvalidate 0.1.2, sortable 0.4.6, stats4 4.2.1, stringi 1.7.8, stringr 1.4.1, this.path 1.0.1, tibble 3.1.8, tidyr 1.2.1, tidyselect 1.2.0, tools 4.2.1, tzdb 0.3.0, utf8 1.2.2, vctrs 0.5.0, withr 2.5.0, xfun 0.34, xml2 1.3.3, xtable 1.8-4, yaml 2.3.6, zlibbioc 1.44.0