# Rapid, interactive visualization of tabular multi-omics data with OmicLoupe

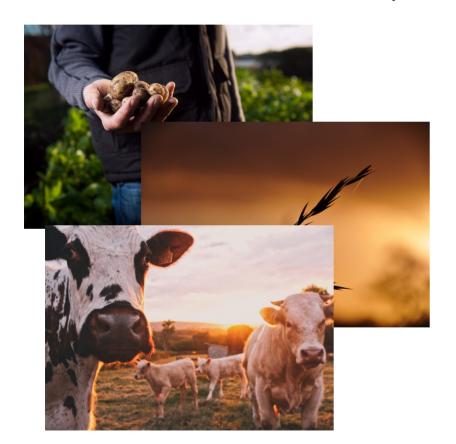
NBIS omics integration course Jakob Willforss, 9<sup>th</sup> October 2020



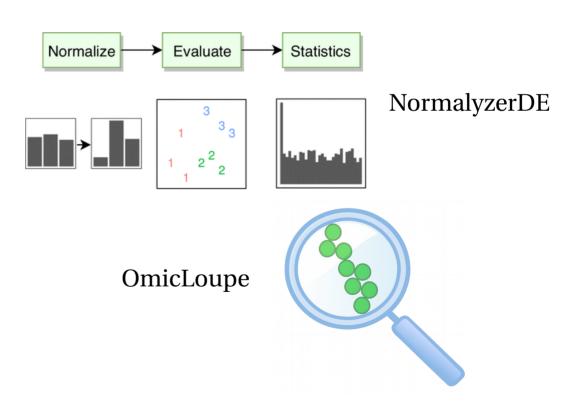


### What I do

Proteomics for biomarker discovery



Improved methods for interpretation of omics



## The plan for this hour

### 1. Lecture

### 2. Demonstration / hands-on

Focus: Integrative study proteomics/transcriptomics

Article | Published: 14 May 2020

## Proteomics of SARS-CoV-2-infected host cells reveals therapy targets

Nature **583**, 469–472(2020) Cite this article

#### RESEARCH ARTICLE

SARS-CoV-2 productively infects human gut enterocytes

Science 03 Jul 2020: Vol. 369, Issue 6499, pp. 50-54 DOI: 10.1126/science.abc1669



## My Ph.D. projects before OmicLoupe

Seasons

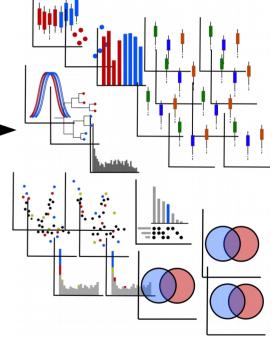














Seasons + validation





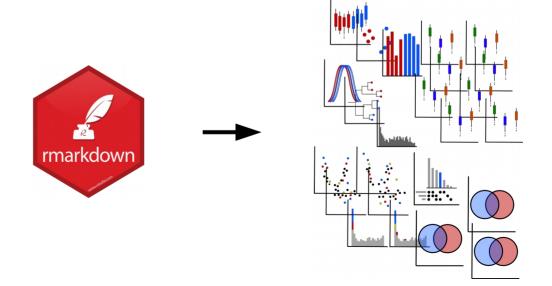




## My Ph.D. projects before OmicLoupe

### Works OK, but:

- Repeated work
- Slow to explore many settings
- Tricky to swiftly inspect details in figures
- Tricky to compare between figures



## With OmicLoupe

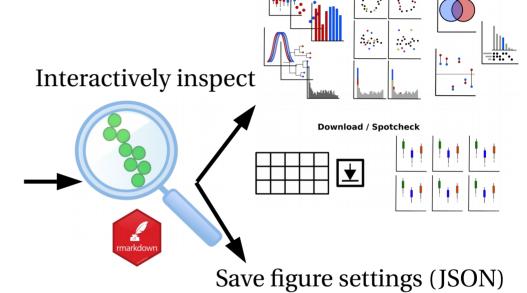
Seasons





**Timepoints** 





**Ouality** 

Statistics

Overlaps

Seasons + validation



## Design philosophy:

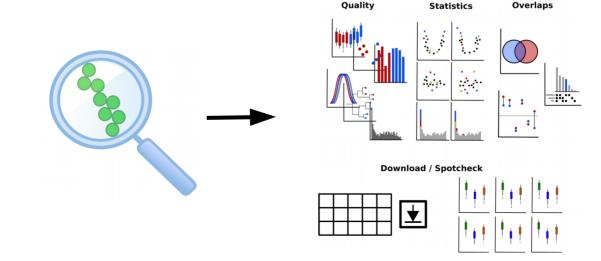
Minimal time & effort in → Maximal output & insight out



Outliers, batch effects, technical trends, biological trends

## What trends are present in each dataset?

## What trends are shared <u>across</u> datasets?



## What is OmicLoupe?

R Shiny + Plot.ly software for rapid, interactive omics-exploration

Tool for **understanding** your single-omics data

Tool for understanding **shared trends** across comparisons/datasets/omics

Reproducible, publication quality visualizations

Can be provided to **end-users** to further explore their data

**Modular**, can easily be extended (maybe with ideas from you!)

### What does it do?

Data generation

Data processing

Statistical analysis

Overlaps

Custom

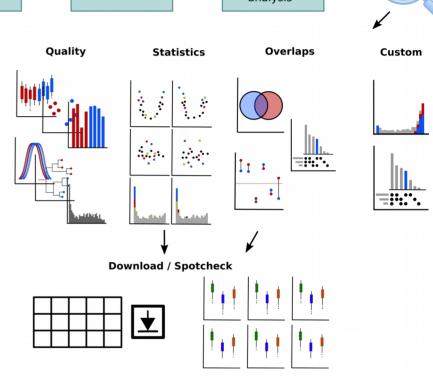
Sample-level visualizations

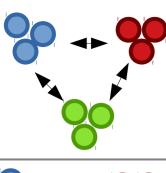
Interactive Qstatistical visualizations

Overlap analysis

Single-feature inspections

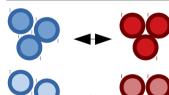
Further custom visualizations



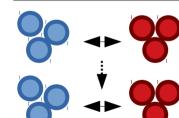


## Comparing trends - Use cases

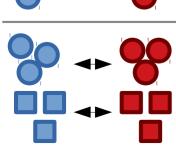
One omic – Multiple comparisons



One dataset – Multiple methods



One study – Multiple repeated samplings



Multiomics ─ Today's focus

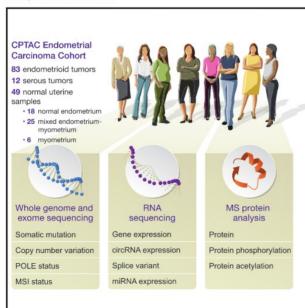
## Visualizing a multi-omics study with OmicLoupe

Resource



### Proteogenomic Characterization of Endometrial Carcinoma

#### **Graphical Abstract**



#### **Authors**

Yongchao Dou, Emily A. Kawaler, Daniel Cui Zhou, ..., Tao Liu, David Fenyö, the Clinical Proteomic Tumor Analysis Consortium

#### Correspondence

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#### In Brief

Proteogenomic analyses of prospectively collected endometrial carcinomas provide insights into the role of underlying molecular pathways and the immune landscape that drive disease.

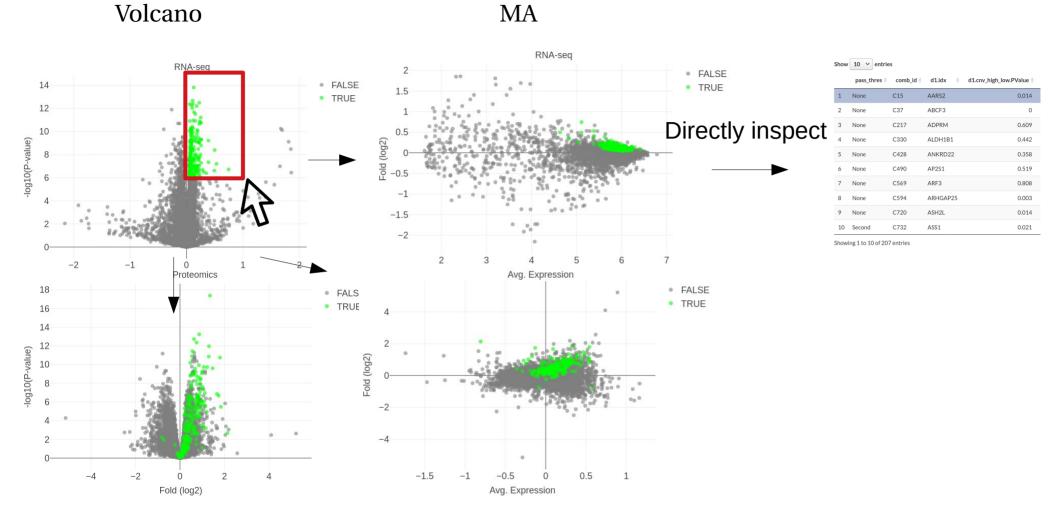
## Quality control (some examples)

PCA Density Boxplot

(Had to remove the figures as they will be used in a publication, sorry)

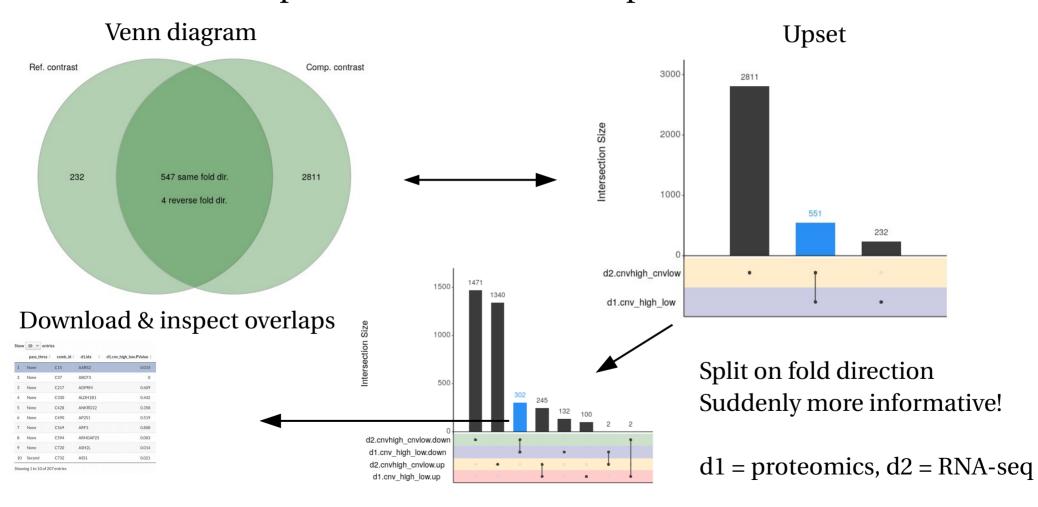
## Statistical comparisons

Volcano MA P-value histogram

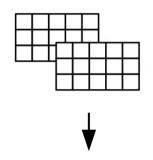


Highlight in figure, see distribution across plots (volcano, MA, p-value histogram)

### High versus low severity tumours (cnvhigh vs cnvlow) In proteomics and transcriptomics



## The workflow



Expression table(s):

RNA-seq, proteomics, metabolomics, microarray ...





NormalyzerDE:

https://github.com/ComputationalProteomics/NormalyzerDE Limma, DESeq2 ... (any other statistical tool)



OmicLoupe explorations

## Sample input

### Data matrix/matrices

(expression data and feature annotations)

### Design matrix/matrices

(sample names and conditions)

### Statistical values for each contrast

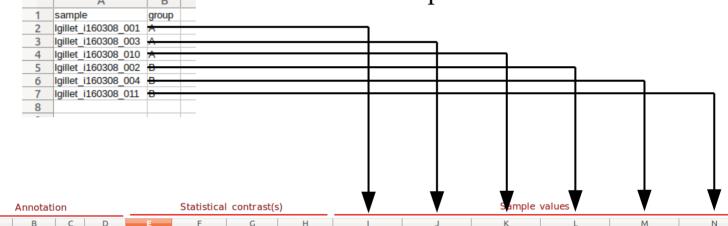
P-values, FDR, log2 fold and average expression

If multiple data matrices: One column with shared IDs

## Sample input: Design- and data matrix

One column matching data matrix samples Columns with sample-conditions

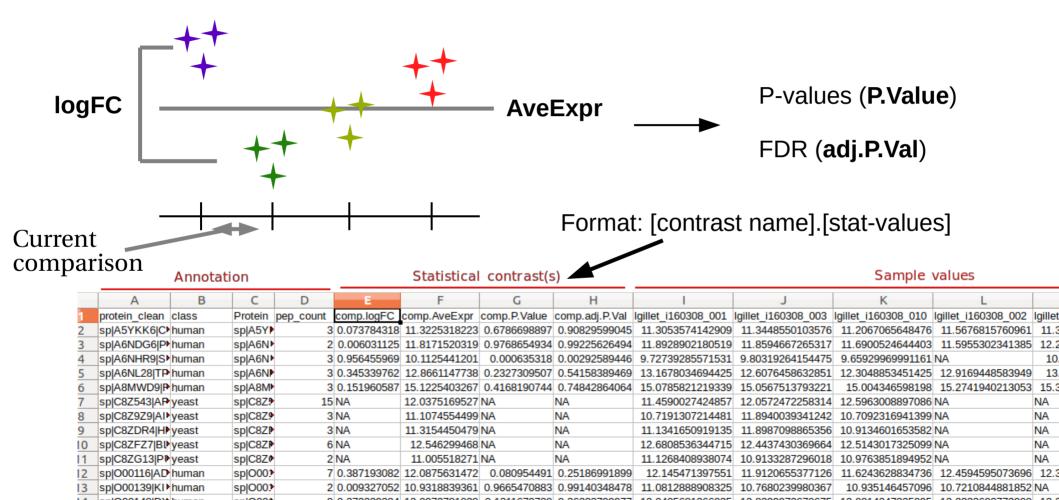
### Design matrix



### Data matrix

	7111101011							<u> </u>	, ,	•	<u> </u>	<u> </u>		
	Α	В	С	D	E	F	G	Н	I	J	K	L	M	N
1	protein_clean	class	Protein	pep_count	comp.logFC	comp.AveExpr	comp.P.Value	comp.adj.P.Val	lgillet_i160308_001	lgillet_i160308_003	lgillet_i160308_010	lgillet_i160308_002	lgillet_i160308_004	lgillet_i160308_011
2	spjA5YKK6jC	human	sp A5Y	3	0.073784318	11.3225318223	0.6786698897	0.90829599045	11.3053574142909	11.3448550103576	11.2067065648476	11.5676815760961	11.3386650815976	11.1719252863554
3	sp A6NDG6 P	human	sp A6N	2	0.006031125	11.8171520319	0.9768654934	0.99225626494	11.8928902180519	11.8594667265317	11.6900524644403	11.5955302341385	12.2099263677684	11.6550461806867
4	sp A6NHR9 S	human	sp A6N	3	0.956455969	10.1125441201	0.000635318	0.00292589446	9.72739285571531	9.80319264154475	9.65929969991161	NA	10.692847883279	10.6799875201249
5	sp A6NL28 TF	human	sp A6NP	3	0.345339762	12.8661147738	0.2327309507	0.54158389469	13.1678034694425	12.6076458632851	12.3048853451425	12.9169448583949	13.502020002742	12.6973891036369
6	spjA8MWD9jF	human	sp A8M	3	0.151960587	15.1225403267	0.4168190744	0.74842864064	15.0785821219339	15.0567513793221	15.004346598198	15.2741940213053	15.3958206314456	14.9255472078366
7	sp C8Z543 AF	yeast	sp C8Z9	15	NA	12.0375169527	NA	NA	11.4590027424857	12.0572472258314	12.5963008897086	NA	NA	NA
8	sp C8Z9Z9 AI	yeast	sp C8Z9	3	NA NA	11.1074554499	NA	NA	10.7191307214481	11.8940039341242	10.7092316941399	NA	NA	NA
9	sp C8ZDR4 H	yeast	sp C8ZP	3	NA NA	11.3154450479	NA	NA	11.1341650919135	11.8987098865356	10.9134601653582	NA	NA	NA
10	sp C8ZFZ7 BI	yeast	sp C8ZP	6	NA	12.546299468	NA	NA	12.6808536344715	12.4437430369664	12.5143017325099	NA	NA	NA
11	sp C8ZG13 PI	yeast	sp C8Z0	2	NA	11.005518271	NA	NA	11.1268408938074	10.9133287296018	10.9763851894952	NA	NA	NA
12	sp O00116 AE	human	sp O00?	7	0.387193082	12.0875631472	0.080954491	0.25186991899	12.145471397551	11.9120655377126	11.6243628834736	12.4594595073696	12.3452771616492	12.0387423955942
13	sp O00139 KI	human	sp 000?	2	0.009327052	10.9318839361	0.9665470883	0.99140348478	11.0812888908325	10.7680239980367	10.935146457096	10.7210844881852	NA	11.1538758462433
14	sp 000148 D>	human	sp 000?	ç	0.273229384	13.0972781038	0.1311679728	0.36223799977	13.0405681366035	12.8399973670675	13.0014247325095	13.2333602773008	13.3602269183308	13.1080911911772
15	sp 000148 D>	human	sp 0003	12	0.093832475	15.9851177832	0.5975302322	0.86227227463	16.0842424333428	15.9020245999128	15.8283376033529	16.2139081723778	15.9557395819064	15.9264543084892
16	sp O00148 D>	ambiguous	sp 000?	2	0.754974214	14.8001969308	0.0324117393	0.11512081996	14.3229700574638	14.138330851045	15.0333208270964	15.5114134323332	NA	14.9949494862226
17	sp O00151 PE	human	sp 000?	2	0.575279174	10.393404225	0.0773638338	0.24287252586	10.1431774462571	9.87562265895916	10.298493808076	11.1054870281953	9.9529521767828	10.9846922315613
18	sp O00154 B	human	sp O00?	5	0.078752393	12.220518744	0.7080232516	0.92004199982	12.0819792418602	12.1861080477041	12.2753403528755	12.4390542786356	11.8578922526316	12.4827382903155
19	splO00159IM	human	splO002	11	0.170384172	12.1034443887	0.3604086284	0.6959318992	11.961578981285	12.1242195565334	11.9689583696402	12.2560119707843	12.3666746037258	11.943222850094

## Sample input: Statistical contrast(s)



## If using multiple tables

#### **Proteomics**

	Α	В	С	D	E	F	G
	gene_symbol	accession	species_names	featureAvg	infected_2h-control_2h_PValue	infected_2h-control_2h_log2FoldChange	infected_2h-control_2h_AdjPVal
2	AFP	P02771	Homo sapiens OX=9606	11.031331403655	0.30969902158687	-0.307359232716664	0.99980939482126
3	FABP1	P07148	Homo sapiens OX=9606	10.954591559394	0.587573151279121	-0.209476148787799	0.99980939482126
1	VIL1	P09327	Homo sapiens OX=9606	10.698272236802	0.416645112976052	-0.191097088944636	0.99980939482126
5	KRT18	P05783	Homo sapiens OX=9606	11.2095110506113	0.633331356317085	-0.132690067072895	0.99980939482126
5		Q9P2E9	Homo sapiens OX=9606	10.4230791312684	0.459488934884819	-0.167897844503694	0.99980939482126
7	ANXA4	P09525	Homo sapiens OX=9606	10.6499881385456	0.82781011607852	-0.050223942845536	0.99980939482126
3	KRT8	P05787	Homo sapiens OX=9606	10.619917799405	0.844472953537969	-0.060373667476336	0.99980939482126
)	MUC13	Q9H3R2	Homo sapiens OX=9606	9.37852849994093	0.831812491341271	0.078448770880204	0.99980939482126
0	CDH17	Q12864	Homo sapiens OX=9606	9.86333520601928	0.518411935369265	-0.142734075113003	0.99980939482126
1	LGALS3	P17931	Homo sapiens OX=9606	10.2031600815392	0.489722341053065	-0.135236521146233	0.99980939482126
2	SERPINA1	P01009	Homo sapiens OX=9606	10.0902246911767	0.55887555105166	-0.136333152852051	0.99980939482126
3	IDH1	O75874	Homo sapiens OX=9606	9.87676816158489	0.184527739086192	-0.358249535050817	0.99980939482126
4	010-400	077010	11 OV 0000	10 4000100000000	0.445000070000014	0.475500000504700	0.000000004004004

### One column need to be in common

	А	R	C	ט	E	F	
	idx	diff_med_24h.logFC	diff_med_24h.AveExpr	diff_med_24h.t	diff_med_24h.P.Value	diff_med_24h.adj.P.Va	
	AAAS	0.028183162940322	5.72297816146273	0.100490173895532	0.921125366636798	0.9917695300499	
	AACS	0.534149879708331	4.45269791483771	1.64919191277447	0.117377841874363	0.4509357956044	
1	AARS	0.199342676608586	4.98757470941001	0.481363034947611	0.636367796688444	0.9279204702562	
,	AASDHPPT	-0.247161129015488	4.83935965937663	-0.695725656217854	0.495959151345674	0.8313211270439	
j	ABCC1	-1.23553413477472	2.70011516424125	-1.65099903822641	0.117004717006166	0.4501955251607	
•	ABCD1	0.152322007640473	0.857479682917489	0.168071491533119	0.86850117891888	0.9917695300499	
}	ABCD3	-0.281670225429062	6.41489681550556	-1.32645373367621	0.202153780378643	0.5642399595980	
)	ABCE1	0.623063181229627	5.35006411905552	2.15407647213151	0.045802950961147	0.296402241764	
0	ABCF1	0.015813080864439	4.88613299788463	0.047478118279708	0.962682923075981	0.9917695300499	
1	ARCE2	_N 12N2266400702N1	2 650721//5/2507	<b>™ 3330ENUU03NN0E3</b>	N 7/12/170570N/2///6	N 00176053NN/00	

RNA-seq

Article | Published: 14 May 2020

## Proteomics of SARS-CoV-2-infected host cells reveals therapy targets

Nature **583**, 469-472(2020) Cite this article

#### RESEARCH ARTICLE

SARS-CoV-2 productively infects human gut enterocytes

© Mart M. Lamers<sup>1,\*</sup>, © Joep Beumer<sup>2,\*</sup>, © Jelte van der Vaart<sup>2,\*</sup>, Kèvin Knoops<sup>3</sup>, Jens Puschhof<sup>2</sup>, Tim I. Breugem<sup>1</sup>, © Ra...

+ See all authors and affiliations

Science 03 Jul 2020: Vol. 369, Issue 6499, pp. 50-54 DOI: 10.1126/science abc1669 Transcriptomics (2 replicates)
24h 72h
Infected
Control
Medium type

Proteomics (3 replicates)

10h

24h

GITHUB: DEMO SERVER: TUTORIAL LINK: github.com/ComputationalProteomics/OmicLoupe http://130.239.81.217:3838/omicloupe

Infected

LINK: quantitativeproteomics.org/analysiscourse/omicloupe

> devtools::install github("ComputationalProteomics/OmicLoupe")

> OmicLoupe::runApp()

Subsequently, click/copy the link "http://127.0.0.1:XXXXX" and run in the web browser

## Thank you for listening!!

Feel free to contact me for further queries!

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