Easy Data Dissemination

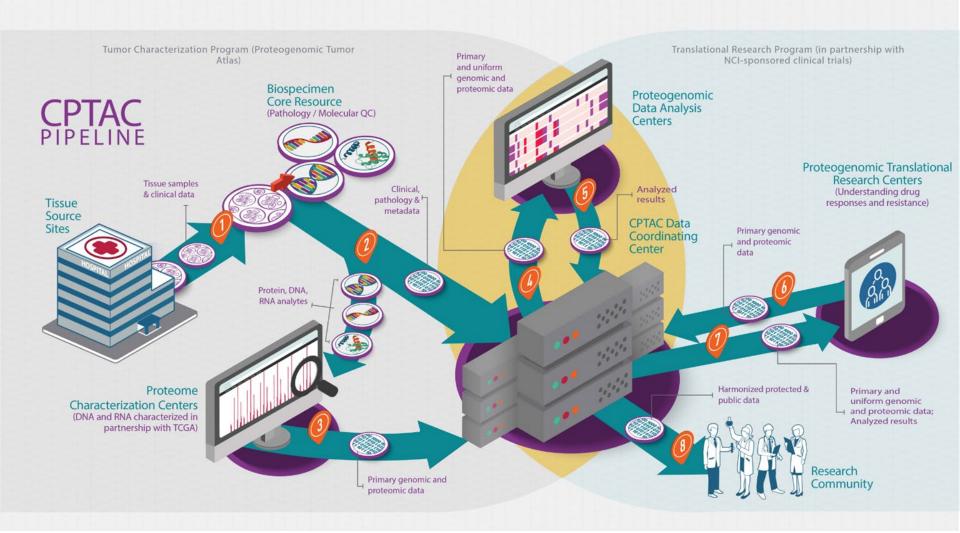
Sam Payne and Caleb Lindgren
Brigham Young University
sam payne@byu.edu

NCI CPTAC award U24 CA210972









Endometrial Cancer

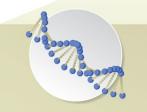
Gather clinical cohort

Multi-omics of tumor samples

CPTAC Endometrial Carcinoma Cohort

- 87 endometrioid tumors
- 13 serous tumors
- 49 normal uterine samples
 - 18 normal endometrium
 - 31 mixed endometriummyometrium





Whole genome and exome sequencing

Somatic mutation

Copy number variation

Structure variation

MSI status



RNA sequencing

Gene expression

Gene fusion

Splice variant

miRNA expression



MS protein analysis

Protein

Protein phosphorylation

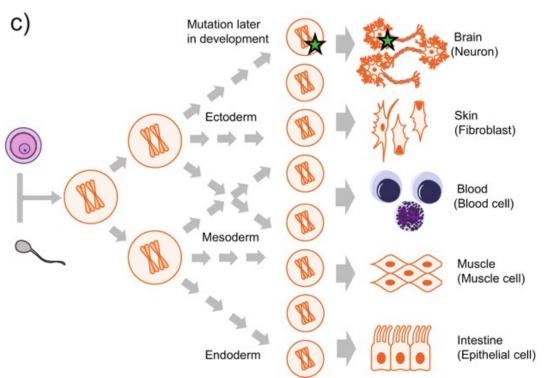
Protein acetylation

Clinical Data

	_	Proteomics_Tumor_Normal	Country	Histologic_Grade_FIGO	Myometrial_invasion_Specify	Histologic_type	Treatment_naive	Tumor_purity	Path_Stage_Primary_Tumor- pT	Path_Sta
Sample_ID		T	Helbert Cheber	5100 mm de 1		Fadamatulald	VEC	Named	-T1- (5100 IA)	
5001	C3L-00006	lumor	United States	FIGO grade 1	under 50 %	Endometrioid	YES	Normal	pT1a (FIGO IA)	
5002	C3L-00008	Tumor	United States	FIGO grade 1	under 50 %	Endometrioid	YES	Normal	pTla (FIGO IA)	
S003	C3L-00032	Tumor	United States	FIGO grade 2	under 50 %	Endometrioid	YES	Normal	pT1a (FIGO IA)	
S005	C3L-00090	Tumor	United States	FIGO grade 2	under 50 %	Endometrioid	YES	Normal	pT1a (FIGO IA)	
S006	C3L-00098	Tumor	United States	NaN	under 50 %	Serous	YES	Normal	pT1a (FIGO IA)	
S007	C3L-00136	Tumor	United States	FIGO grade 1	under 50 %	Endometrioid	YES	Normal	pT1a (FIGO IA)	
S008	C3L-00137	Tumor	Other_specify	FIGO grade 2	under 50 %	Endometrioid	YES	Normal	pTla (FIGO IA)	
S009	C3L-00139	Tumor	United States	NaN	50 % or more	Serous	YES	Normal	pT3a (FIGO IIIA)	
S010	C3L-00143	Tumor	United States	FIGO grade 1	Not identified	Endometrioid	YES	Normal	pT1 (FIGO I)	
5011	C3L-00145	Tumor	United States	FIGO grade 1	under 50 %	Endometrioid	YES	Normal	nTla (FIGO IA)	

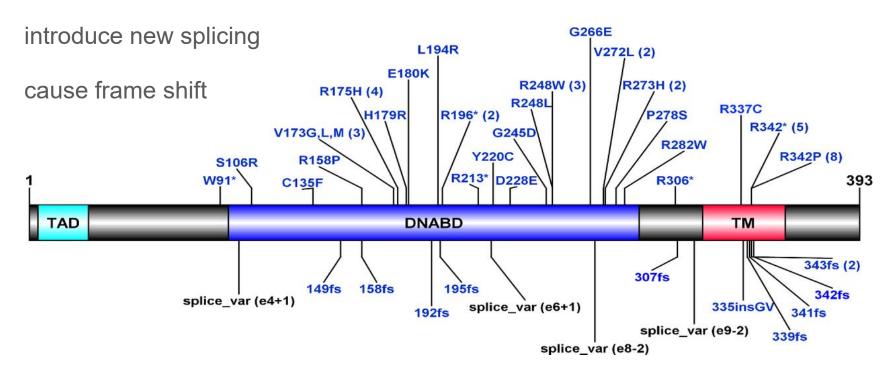
Genomics Data

Somatic mutation - changes to DNA after embryo, present only in some cells



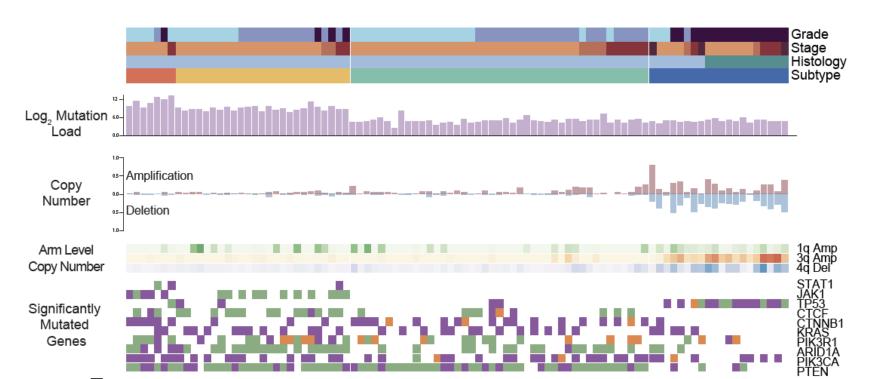
Single Nucleotide Mutations

change an amino acid



Exploring Data

Look at clinical and genomics data to find mutations that correlate with histology

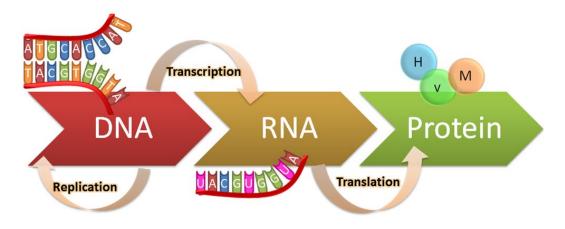


Mutation and clinical data

	Histologic_type	TP53_Mutation	TP53_Location	TP53_Mutation_Status	Sample_Status
Sample_ID					
S001	Endometrioid	Missense_Mutation	p.R248W	Single_mutation	Tumor
S002	Endometrioid	Wildtype_Tumor	No_mutation	Wildtype_Tumor	Tumor
S003	Endometrioid	Wildtype_Tumor	No_mutation	Wildtype_Tumor	Tumor
S005	Endometrioid	Wildtype_Tumor	No_mutation	Wildtype_Tumor	Tumor
S006	Serous	Missense_Mutation	p.S241C	Single_mutation	Tumor
S007	Endometrioid	Wildtype_Tumor	No_mutation	Wildtype_Tumor	Tumor
S008	Endometrioid	Missense_Mutation	p.S127Y	Single_mutation	Tumor
S009	Serous	Nonsense_Mutation	p.R196*	Single_mutation	Tumor
S010	Endometrioid	Wildtype_Tumor	No_mutation	Wildtype_Tumor	Tumor
S011	Endometrioid	Wildtype_Tumor	No_mutation	Wildtype_Tumor	Tumor
S012	Endometrioid	Wildtype_Tumor	No_mutation	Wildtype_Tumor	Tumor
S014	Endometrioid	Wildtype_Tumor	No_mutation	Wildtype_Tumor	Tumor

Multi-omics data

DNA, RNA, Proteins and cellular state



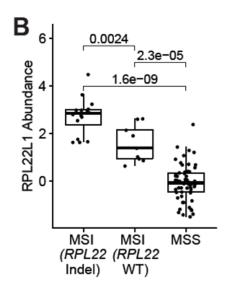
DNA, RNA measured with 'sequencing'

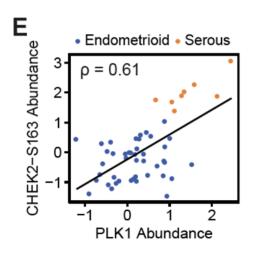
Protein measured with mass spectrometry

	CTNNB1	JAK1	PIK3CA	PTEN	TP53
Sample_ID					
S001	0.80300	-0.38200	-0.4500	-0.5260	0.295000
S002	0.42600	-0.39300	-0.2150	-0.8300	0.277000
S003	0.00891	-0.09580	-0.1780	-0.9410	-0.871000
S005	0.78800	-0.62700	-0.3920	0.7300	-0.343000
S006	-0.40200	0.14100	0.2870	-0.3790	3.010000
S007	1.43000	-0.22900	-0.3490	0.0293	-0.148000
S008	0.97100	-0.54800	-0.3170	-1.0100	0.441000
S009	-0.25200	-0.36000	0.2820	0.1300	-1.220000
S010	0.00979	0.30400	-0.0714	0.3900	-0.082500
6011	0 50100	0.44100	0 0640	0 0201	A 101AAA

Multi-omics data integration

- B. Protein abundance, mutation status, cancer subtype
- E. Protein abundance, phosphorylation abundance, histology





cptac package

Python package for Data as a Service

data is 'in code, not an excel file'

```
Below are the dataframes contained in this dataset:
        acetylproteomics
                Dimensions: (144, 10862)
        circular RNA
                Dimensions: (109, 4945)
        clinical
                Dimensions: (144, 26)
        CNV
                Dimensions: (95, 28057)
        derived molecular
                Dimensions: (144, 125)
        experimental setup
                Dimensions: (144, 26)
        miRNA
                Dimensions: (99, 2337)
        phosphoproteomics
                Dimensions: (144, 73212)
        phosphoproteomics gene
                Dimensions: (144, 8466)
        proteomics
                Dimensions: (144, 10999)
        somatic mutation
                Dimensions: (52560, 3)
        somatic mutation binary
                Dimensions: (95, 51559)
        transcriptomics
                Dimensions: (109, 28057)
```

Interactive demo of a notebook

View demo at

https://colab.research.google.com/drive/17SMHMDeOUMdkJAd2Xc3Hgvvyqv_W EW4r

How to work with imaging data

imaging features as a table (nucleus_major_axis, nucleus_aspect_ratio, etc.) correlating imaging features with clinical, omics, etc.



Journal of Digital Imaging

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Predicting Deletion of Chromosomal Arms 1p/19q in Low-Grade Gliomas from MR Images Using Machine Intelligence