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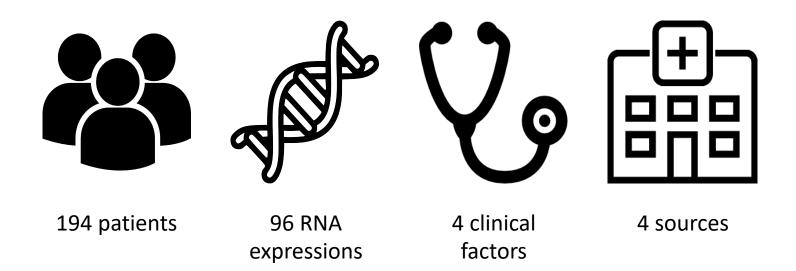
Supv.: Prof. Ieva and Prof. Ceri



# Identification and validation of epigenetic lesions in refractory lymphomas

#### Aim and Dataset

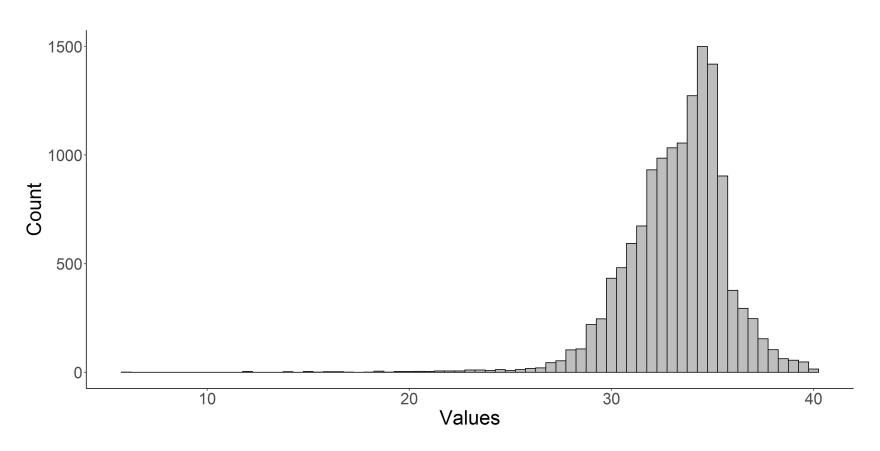
Upfront identification of refractory classical Hodgkin lymphoma (cHL) patients



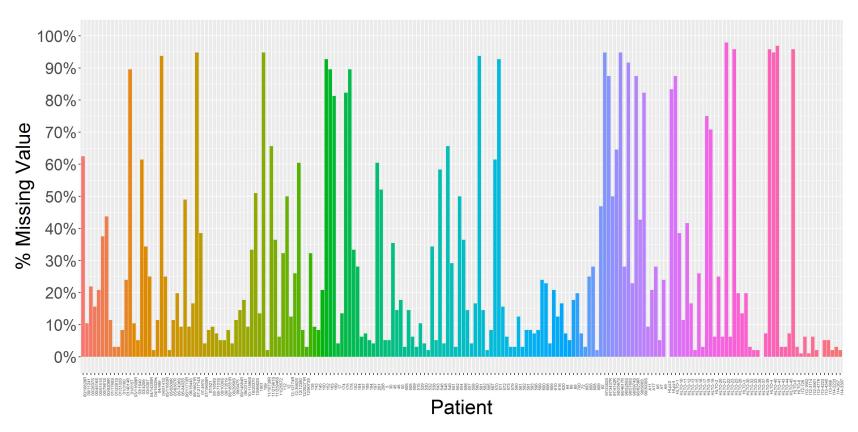
#### Work schedule on Raw data



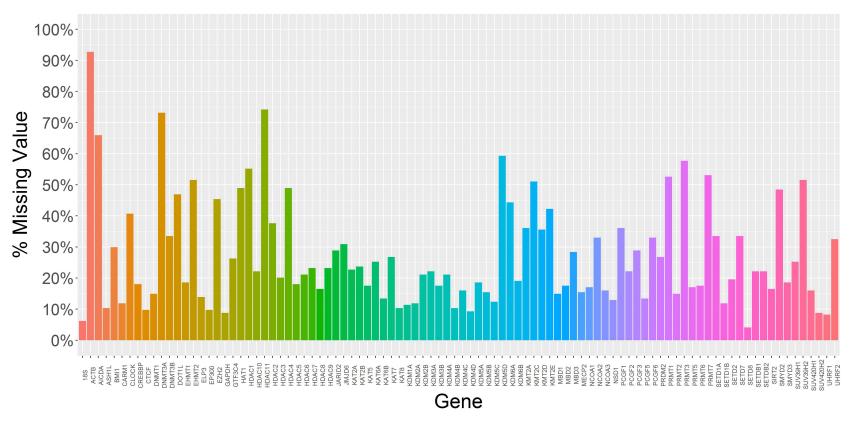
# Preliminary Data Analysis: Data Distribution



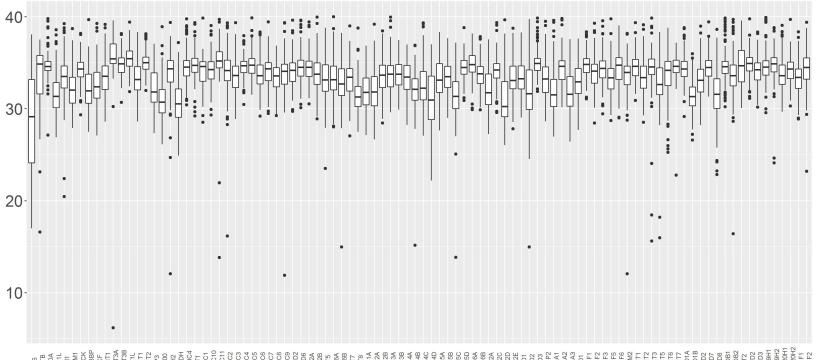
# Preliminary Data Analysis: Missing data for Patient



### Preliminary Data Analysis: Missing data for Gene



### Preliminary Data Analysis: Outliers



188 AGTB BMH1 BMH1 BMH1 BMH1 CARM1 CARM1 CARM1 CLOCK CREBEP CATTA CAT

Genes

#### Preliminary Data Analysis

#### Missing data

Missing analysis

Imputation procedure

#### **Outliers**

Outlier analysis

Removal of significant outliers

#### Data Cleaning:

Identification of Normalization Factor (NF)

#### Literature

- GAPDH
- Literature on Hodgkin Lyphoma (GAPDH and ACTB)

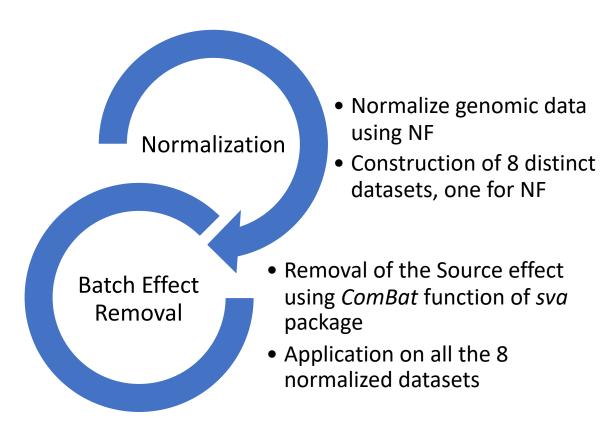
# Original Data

- Pairwise Comparison Approach (GeNorm)
- Model-based Approach (NormFinder)
- Pairwise Correlation Approach (BestKeeper)

# Combined Ranking

- Pairwise Comparison Approach (GeNorm)
- Model-based Approach (NormFinder)
- Pairwise Correlation Approach (BestKeeper)

### Data Cleaning: Normalization and Batch Effect Removal



#### Classifier Construction

#### Best classifier

- NF obtained by GeNorm method
- Classification method: CART

	Patient	Genes
1	194	96

	R	$\mathbf{S}$
R	26	31
$\mathbf{S}$	30	101

	Accuracy	Kappa	AUC
1	0.6546	0.2422	0.7116

	TPR	FPR	Precision	F1
$\mathbf{R}$	0.4643	0.2348	0.4561	0.4602
$_{\rm S}$	0.7652	0.5357	0.7710	0.7681

#### Classifier Constructon with Source

#### Best Classifier

• NF: GAPDH (literature)

• Classification method: CART

	Patient	Genes
1	194	96

	R	$\mathbf{S}$
R	27	18
$\mathbf{S}$	34	115

	Accuracy	Kappa	AUC
1	0.7320	0.3308	0.6739

	TPR	FPR	Precision	F1
R	0.4426	0.1353	0.6000	0.5094
$\mathbf{S}$	0.8647	0.5574	0.7718	0.8156

Classification methods dealing with group effect

# Work schedule: *Missing values issue*

Preliminary data analysis

Data cleaning

Classifier construction

#### Imputation

#### Random uniform values in [40,45]

- Data concentration at high values
- Generation of fictitious outliers

#### Fixed value at 40

- Data concentration at 40
- Abnormal trends become regular

#### Mean of gene

- Data concentration in central band
- Variance reduction

#### Median of gene

- Data concentration in central band
- Variance reduction

#### Multiple Imputation Method (mice package)

- Imputation with observed data (more realistic)
  - Influenced by significant low outliers

#### Classifier Construction

# Best classifier

- Imputation: *mice* package
- NF obtained by BestKeeper method on Final Ranking
- Classification method: RandomForest

	Patient	Genes
1	194	96

	R	S
R	24	8
$\mathbf{S}$	37	125

	Accuracy	Kappa	AUC
1	0.7680	0.3825	0.7309

	TPR	FPR	Precision	F1
R	0.3934	0.0602	0.7500	0.5161
S	0.9398	0.6066	0.7716	0.8475

#### Classifier Construction with Source

# Best classifier

- Imputation: mice package
- NF obtained by BestKeeper method
- Classification method: RandomForest

	Patient	Genes
1	194	96

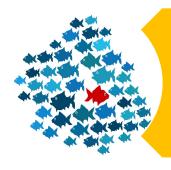
	R	S
$\mathbf{R}$	25	9
$\mathbf{S}$	36	124

	Accuracy	Kappa	AUC
1	0.7680	0.3887	0.7244

	TPR	FPR	Precision	F1
R	0.4098	0.0677	0.7353	0.5263
$_{\rm S}$	0.9323	0.5902	0.7750	0.8464

Classification methods dealing with group effect

#### Next Steps



# Removal of significant outliers



Classification methods dealing with group effect