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# Handling confounding factors in analyzing the transcriptomic data from Chornobyl tree frogs

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# I. Context



Nuclear power plant accident in Chornobyl, Ukraine (1986): release of ionizing radiation and environmental contamination





Collection of tree frog population samples

- 87 males from 8 sites, including 6 within the Chornobyl Exclusion Zone (CEZ)
- Exposed to various levels of contamination
- Omics and non-omics data extraction:
  - Dosimetry: total dose rate with internal and external contributions
  - Age, phenotype (mass, dimensions)
  - Genomics (DNA-seq)



Confounding factor for the transcriptomic data analysis



A principal component analysis (PCA) performed on the RNA-seq data reveals the association with the site. The site factor appears to have a stronger role than the dose in explaining the variability in the transcriptomic data.

Sampling locations of Eastern tree frog Hyla orientalis populations in the exclusion zone in 2018[1,2]

Transcriptomics (RNA-seq)

Stad Proteomics

# **Objective:** Identify specific molecular signatures of exposure to better understand the effects of low-dose radiation

#### Methodological challenge

 $\rightarrow$  Handling factors confounding the analysis of the effect of the dose

II. Strategies for batch effect mitigation	When to use	How to use in R	Principle	Algorithm
Residualization	Pre-processing step	Native stats package	Removes linear relationship between variables and batch	<ul> <li>For each variable:</li> <li>1. Fit a linear regression model on the variable using the batch factor as a regressor</li> <li>2. Extract the residual part</li> </ul>
How to handle batch effects (BEs)?	Pre-processing step [3]	sva package	Corrects data distribution to fit a batch-free distribution	1. Estimate the parameters of the negative binomial regression model $y_{gij} \sim NB(\mu_{gij}, \phi_{gi})$ For gene $g$ , sample $j$ , batch $i$ : $\log \mu_{gij} = \alpha_g + X_j \beta_g + \gamma_{gi} + \log N_j$ 2. Compute the batch-corrected distributions $Adjust data to the batch-corrected distributions$ by quantile mappingFor gene $g$ , sample $j$ , batch $i$ : $\log \mu_{gij} = \alpha_g + X_j \beta_g + \gamma_{gi} + \log N_j$ $average levelof expression$
BES MINT PCA	Integrated method [4]	mixOmics package	Within-group centering and scaling followed by PCA	1. Data is centered and scaled within groups 2. PCA algorithm is performed (based on SVD matrix diagonalization) $X_1$ $X_1$ $X_2$ $X_1$ $X_1$ $X_2$ $X_1$ $X_2$ $X_1$ $X_2$ $X_2$ $X_2$ $X_1$ $X_2$ $X_2$ $X_2$ $X_2$ $X_1$ $X_2$ $X_2$ $X_1$ $X_2$ $X_2$ $X_1$ $X_2$ $X_1$ $X_2$ $X_1$ $X_2$ $X_2$ $X_1$ $X_2$ $X_2$ $X_1$ $X_2$ $X_2$ $X_1$ $X_1$ $X_1$ $X_2$ $X_1$ $X_2$ $X_1$ $X_2$ $X_1$ $X_1$ $X_2$ $X_1$ $X_1$ $X_1$ $X_1$ $X_1$ $X_1$ $X_2$ $X_1$

Batch effect mitigation is a common harmonization problem when integrating data obtained across different experiments, individuals or technologies. Methods either account for BEs, by including the batch as a covariate, or correct them by removing the unwanted variation in a pre-treatment step. We applied PCA on the corrected count matrices to compare performances with those of MINT PCA.

the

was



Visualization of results through PCA shows no substantial difference between all three methods: the



#### **IV. Controlling for the genetic diversity**



**Results** 

methods

Residualization, ComBat\_seq, and MINT PCA yielded qualitatively satisfactory results: monitoring the effect of genetic distance revealed an association between increased dose rate and variation in gene expression.



#### **Treating genetic diversity as a batch effect**

To take genetic diversity into account, hierarchical clustering was performed on the inter-individual genetic distance matrix. This hierarchical clustering enabled genetic diversity to be summarized in a 3level categorical variable. The groups obtained reflect the coherence between genetic structure and geography and are consistent with the migrations observed between sites within the exclusion zone [5]. The methods described previously were applied to the RNA-seq count data to correct/account for the genetic diversity grouping.





#### The site: an obscure variable

Severe confoundedness between

The site effect is particularly complex to

remove as it is severely confounded with

principal components and dose rate was

computed to test the preservation of dose-

related biological information. This metric

to

dose rate. The correlation between

compare

site and dose: a challenge

used

quantitatively.

This 8-level categorical variable encapsulates information on site-tovariations such as local site environment and food sources, day of capture, and genetic diversity between sampled populations.

### **V. Conclusion and perspectives**

#### **Limits of BE mitigation methods:**

Lack of reliability when applied to a condition-confounded batch.

- ComBat\_seq's condition input parameter allows the preservation of condition-related variation but only accepts qualitative variables.
- To reduce the effect of the site, we focused on taking genetic diversity into account as a batch factor, allowing us to extract the variation in expression associated with exposure.

#### **Perspectives include:**

Multi-modal omics dataset integration (RNA-seq, DNA-seq, proteomics).

#### References

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- Icônes : Flaticon.com (graphistes : Freepik, surang, Yogi Aprelliyanto, lapivee)

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