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# Hierarchical classification of ncRNAs in small RNA seq data based on block patterns

35<sup>th</sup> Winterseminar Bled

Tobias Hagemann

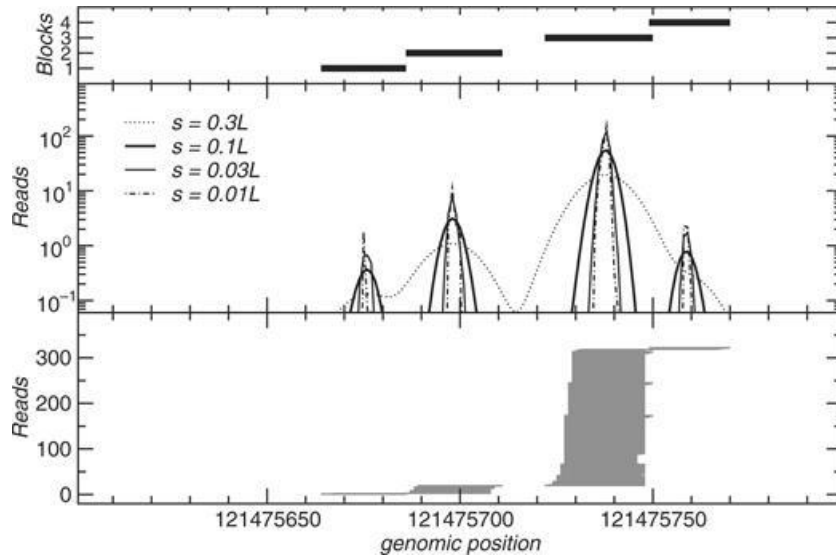
Bioinf Leipzig

12.02.2020



# Basics

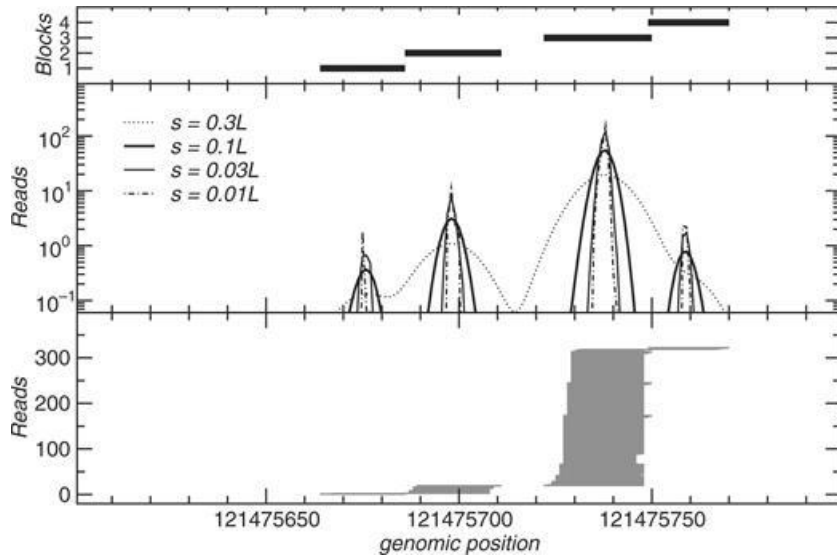
## BLOCKBUSTER 1)



1) Langenberger et al. (2009)

# Basics

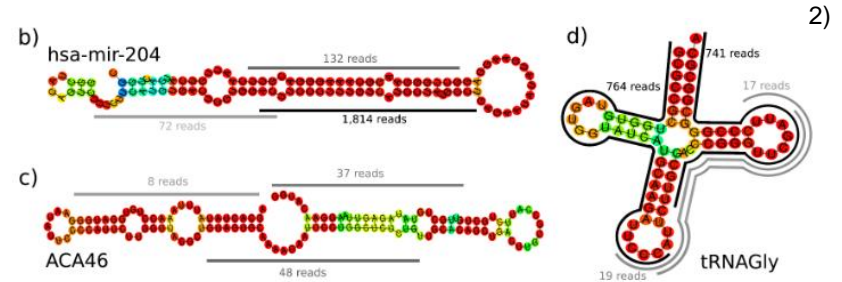
## BLOCKBUSTER 1)



1) Langenberger et al. (2009)

## DARIO 2) 3)

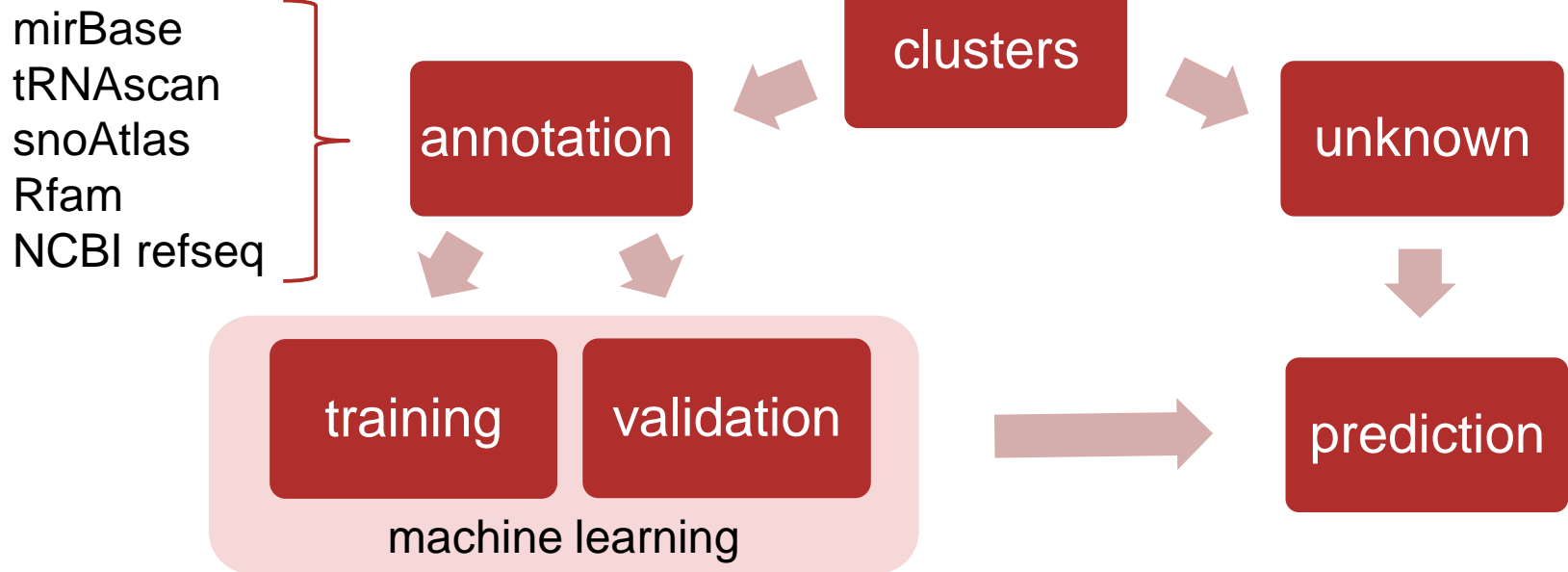
- Train and classify transcribed regions
- Use block patterns
- For tRNA, miRNA, snoRNA



2) Langenberger et al. (2010)

3) Fasold et al. (2011)

## Basic idea



# Learning on bam files



## Account for differences in

- Mapping
- Sequencing
- Species
- Tissue
- ...



# Learning on bam files



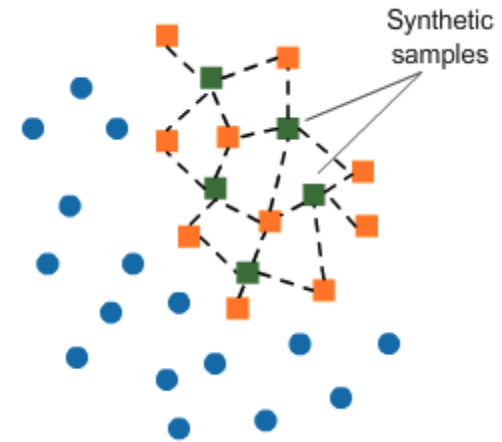
Account for differences in

- Mapping
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- ...

Small sample size & class imbalance

# Synthetic minority oversampling technique (SMOTE)

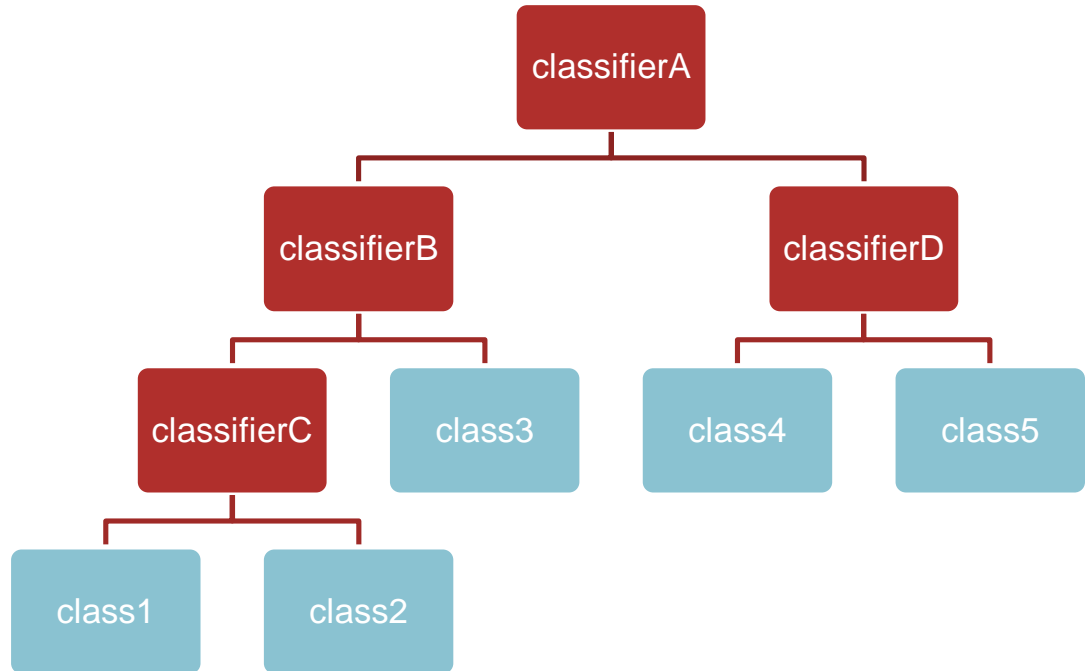
Class	sample size before SMOTE	Sample size after SMOTE
tRNA	326 (36,2%)	326
snoRNA	218 (24,2%)	326
mRNA	210 (23,3%)	326
miRNA	68 (7,5%)	326
rRNA	57 (6,3%)	326
$\Sigma$	879 (97,5%)	1630



<https://towardsdatascience.com/super-bowl-prediction-model-99048f366fed>

# Hierarchical classification

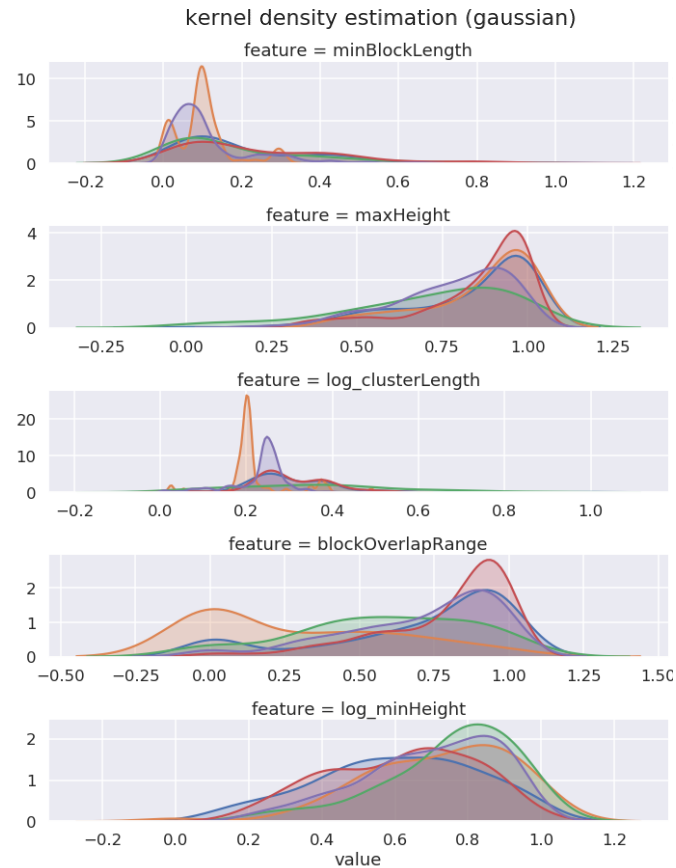
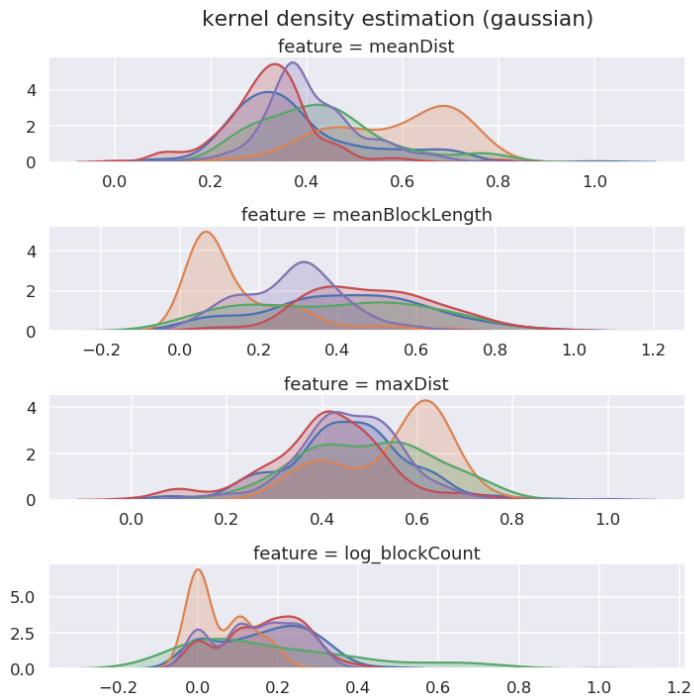
- “Local classifier per parent node“ approach <sup>4)</sup>
- Feature selection and clustering at every parent node



4) Silla & Freitas (2001)

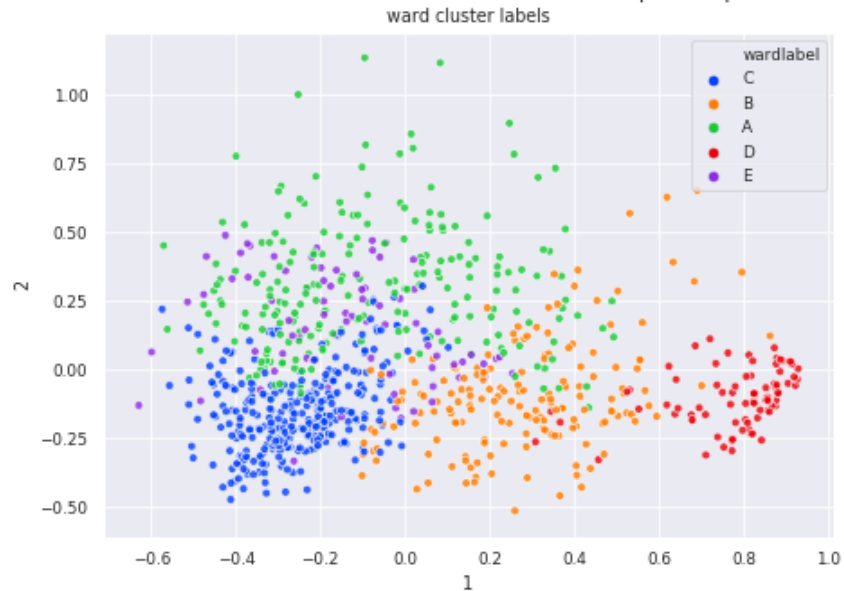


# Block feature analysis



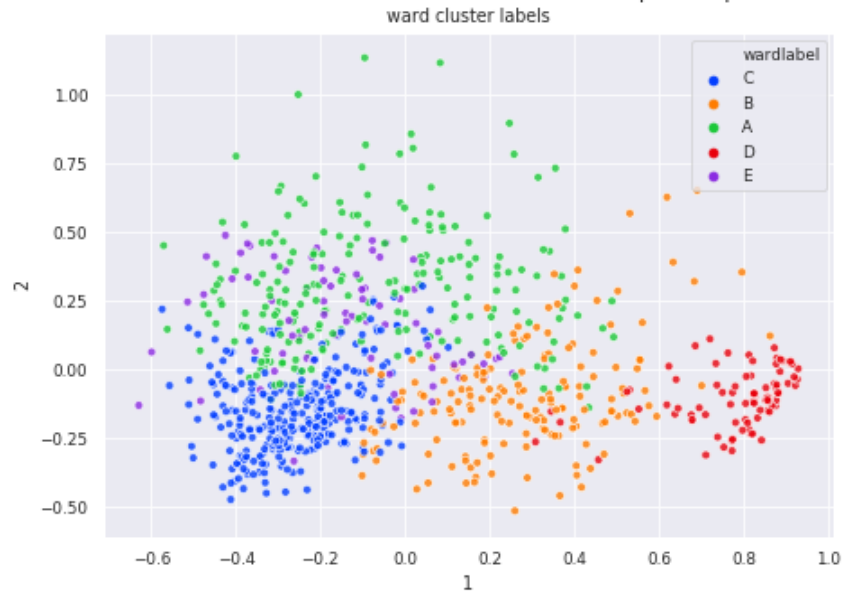
# Block feature analysis

First Two Principle Components on Block Features (explaining 70% variance)



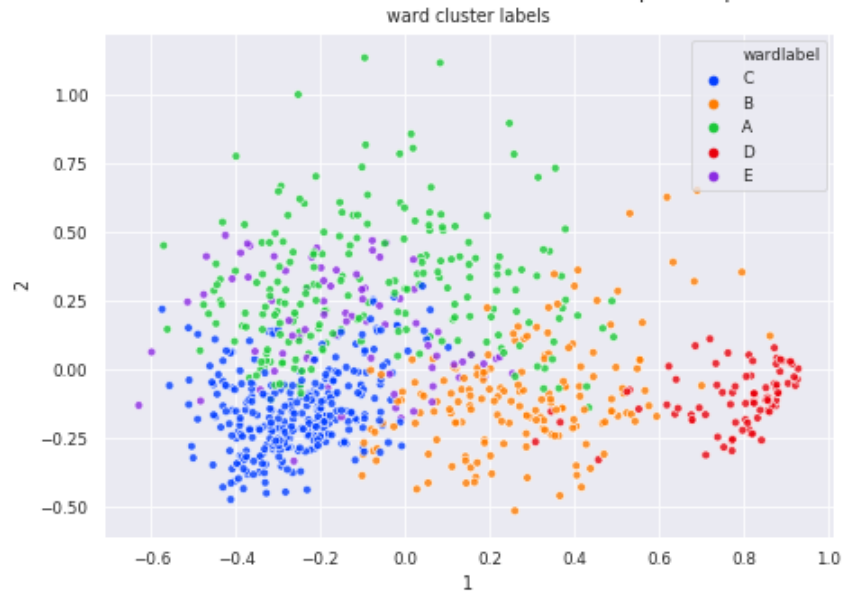
# Block feature analysis

First Two Principle Components on Block Features (explaining 70% variance)



# Block feature analysis

First Two Principle Components on Block Features (explaining 70% variance)



# Block feature engineering on cluster level

Feature type based on ...	Possible Features	
Sequence	<ul style="list-style-type: none"><li>• GC content/block</li><li>• K-tuple composition/block (Shannon entropy)</li></ul>	} Novel structures but Less conservation
Mapping	<ul style="list-style-type: none"><li>• Mismatches/block</li><li>• Average quality/block</li></ul>	
Structure	<ul style="list-style-type: none"><li>• Mfe</li><li>• Self-containment</li><li>• Base pair entropy</li><li>• Accessibility (at or in between blocks)</li><li>• Hairpins/block</li></ul>	} Better performance but overfitting

## Sources

- 1) Langenberger, David, et al. "Evidence for human microRNA-offset RNAs in small RNA sequencing data." *Bioinformatics* 25.18 (2009): 2298-2301
- 2) Langenberger, David, et al. "Identification and classification of small RNAs in transcriptome sequence data." *Biocomputing 2010*. 2010. 80-87.
- 3) Fasold, Mario, et al. "DARIO: a ncRNA detection and analysis tool for next-generation sequencing experiments." *Nucleic acids research* 39.suppl\_2 (2011): W112-W117.
- 4) Silla, Carlos N., and Alex A. Freitas. "A survey of hierarchical classification across different application domains." *Data Mining and Knowledge Discovery* 22.1-2 (2011): 31-72.



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**THANKS TO ...**

Stephanie Kehr

Peter Stadler

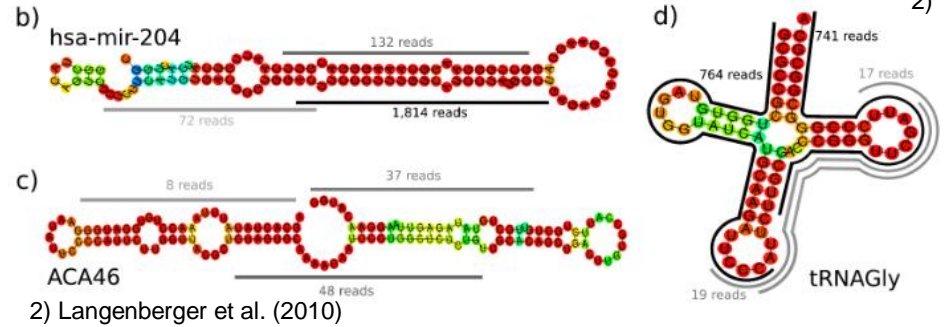
Viewpoint Hunters

You!



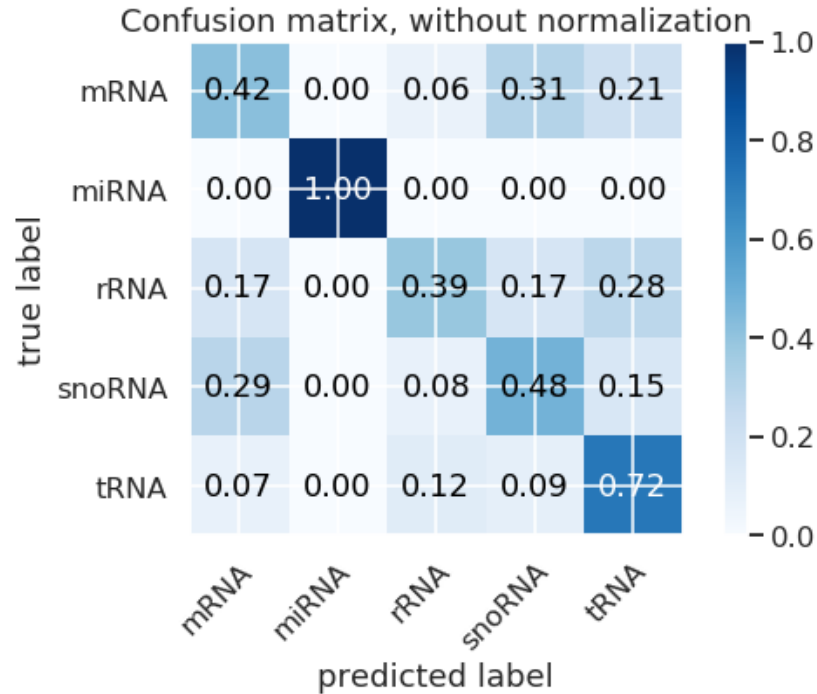
# DARIO performance

class	PPV	Recall
tRNA	0,932	0,918
miRNA	0,860	0,633
snoRNA	0,819	0,694

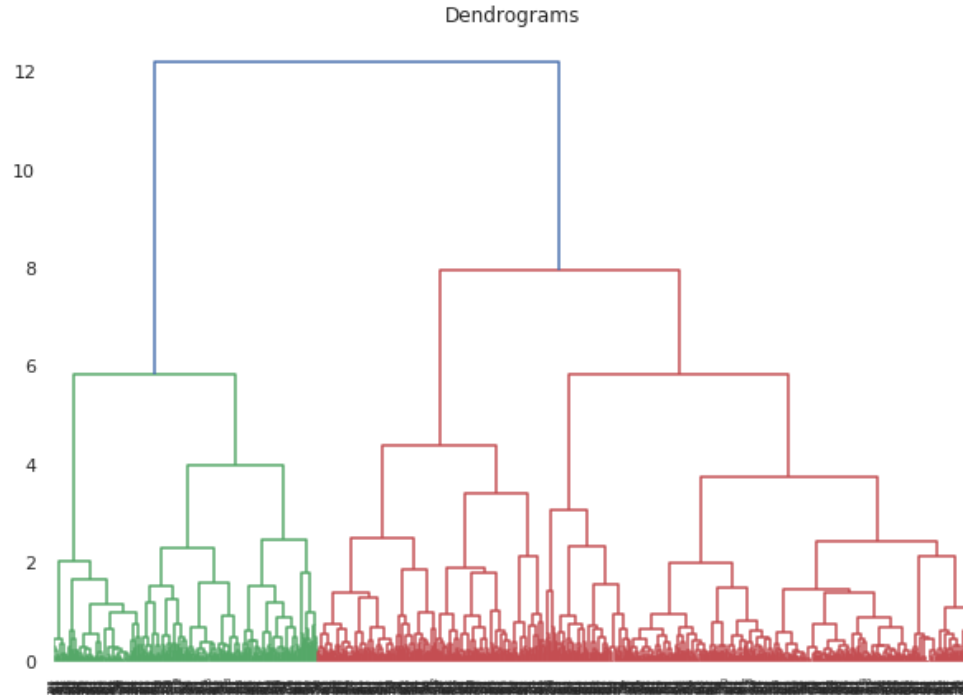




# Random forest prediction on 5 classes



# Ward clustering



# Block features (DARIO)

