# A Computational Platform for Gene Expression Analysis

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## Outline

- 1 Introduction
  - Domain Problem
  - Motivation and Objectives
- 2 Developed Solution
  - Overview
  - RNA-Seq Analysis Pipeline
  - RBP Analysis Pipeline (PBS Finder)
  - Integration
- 3 Case Studies
  - RNA-Seq Analysis Pipeline
  - RBP Analysis Pipeline (PBS Finder)
- 4 Conclusions
  - Objective Fulfilment
  - Future Work

## Domain Problem I

#### Introduction

Molecular biology is a young field of study, with a lot of unknowns and partial knowledge.

- Studying gene expression is crucial to understand the mechanisms that control living organisms.
- We focused on two different areas:
  - differential expression analysis;
  - RNA-binding protein (RBP) discovery and analysis.

## Domain Problem II

#### Introduction

## Three distinct problems:

- Read alignment against a reference genome and differential expression analysis on the aligned data.
- RBP discovery, analysis and information enrichment.
- Further result analysis using data mining techniques.

#### Introduction

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#### Gather information

Information should be contextually aggregated, allowing for quick access of relevant information.

## Overview

### Developed Solution

■ Two distinct problems warrant two different solutions.

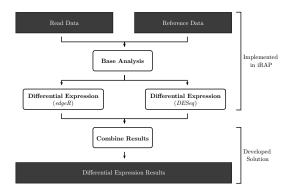
■ The developed system should be available anywhere, through the internet.

■ The system should be as modular as possible, to allow future extensions.

## RNA-Seq Analysis Pipeline I

### Developed Solution

- Uses iRAP as the analysis pipeline.
- Conducts multiple differential expression analyses with different tools.
- Combines results from multiple tools.



# RNA-Seq Analysis Pipeline II

### Developed Solution

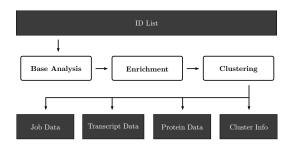
#### Additional features:

- Access to iRAP's web reports and gene browser, along with result visualization in the web interface.
- Synchronization with Ensembl's reference genome repositories.
- Graphical job configuration.
- Possibility to easily include other differential expression tools.

# RBP Analysis Pipeline (PBS Finder) I

Developed Solution

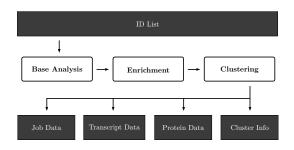
 Uses Ensembl and NCBI to identify gene species, obtain basic information and extract genetic sequences (5' UTR, 3' UTR, 3' UTR downstream).



## RBP Analysis Pipeline (PBS Finder) II

### Developed Solution

- Uses RBPDB to discovery RNA binding proteins based on the obtained sequences.
- Uses UniProt to enrich the obtained results and performs clustering analysis on those results.



# RBP Analysis Pipeline (PBS Finder) III

Developed Solution

## Clustering analysis:

- Uses k-medoids and hierarchical clustering, both with Jaccard and binary distance matrices.
- Executes every possible combination of clustering setups (alternates algorithms, distance matrices, used features, etc.).
- Results are filtered (acceptable solutions must have a minimum percentage of entries per cluster, clusters must have defining features, etc.).
- Solution quality internally determined based on the average silhouette.

# RBP Analysis Pipeline (PBS Finder) IV

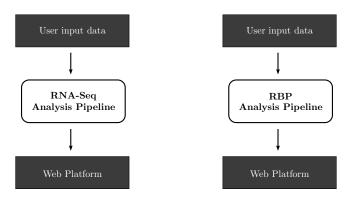
Developed Solution

#### Additional features:

- User account and job management system.
- References to external platforms with relevant information based on context.
- Support for multiple identifier notations (Ensembl, Entrez, RefSeq and GenBank).
- Visualization of defining features for each cluster.
- Job completed notification system.

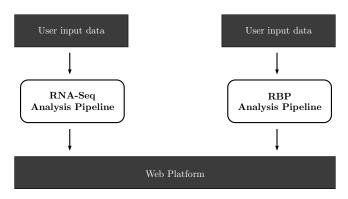
### Developed Solution

While focusing on aggregation and quick access to information, does it make sense to separate the results into two different platforms?



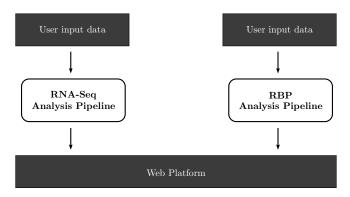
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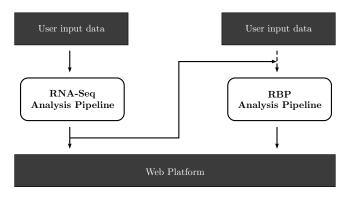
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A list of differentially expressed genes is not very useful without further information about those genes. Does it make sense for a user to launch a new gene enrichment task by hand?



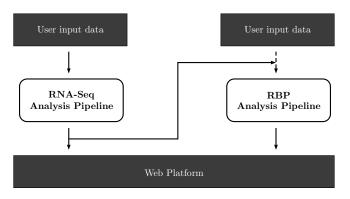
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### Developed Solution

A fully integrated solution: the analysis pipelines can be used separately or automatically executed in sequence; result visualization for both pipelines is isolated.



# RNA-Seq Analysis Pipeline I

Case Studies

## Objective

• Ascertain if combining the results of multiple tools has impact on the set of differentially expressed genes.

#### Data set

- Reproduction of ArrayExpress experiment E-GEOD-48829 (Escherichia coli).
- Reference genome obtained from Ensembl Genomes and read data obtained from ENA Sequence Read Archive.

# RNA-Seq Analysis Pipeline II

Case Studies

Results (number of differentially expressed genes)

	Raw results	$Filtered\ results$	$Combined\ results$
edgeR	4494	386	191
DESeq	4494	204	191

#### Conclusions

- Combining results impacts the final differentially expressed gene list by reducing its size.
- The combined results will hopefully give researchers an higher confidence in the experimental results.

# RBP Analysis Pipeline (PBS Finder) I

Case Studies

## Objectives

- Assess the general usefulness of PBS Finder.
- Compare PBS Finder with the existing techniques of manual analysis.
- Assess the impact of differences in hardware performance in the overall performance of the platform.

#### Data set

■ 23 genes from the *RhoGTPase* family (*Rattus norvegicus*) provided by IBMC.

# RBP Analysis Pipeline (PBS Finder) II

Case Studies

Results (expert estimation of 30 minutes per gene analysed)

$Number\ of\ IDs$	Machine 1	Machine 2	$Manual\ method$
100	$9m \ 56s$	$11m \ 1s$	$\approx 50h$
500	$41m \ 47s$	55m 51s	$\approx 250h$
900	$1h \ 33m \ 32s$	$2h\ 7m\ 4s$	$\approx 450h$

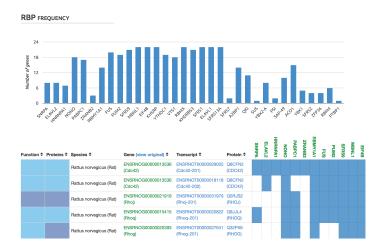
#### Conclusions

- PBS Finder can reproduce the results an expert would get.
- Months worth of an expert's manual work can be accomplished in a few hours.
- While hardware performance has a significant impact on analysis time, the platform achieves satisfactory performance on personal computer-level hardware.

# RBP Analysis Pipeline (PBS Finder) III

#### Case Studies

## Results (viewed in PBS Finder)



# Objective Fulfilment

#### Conclusions

■ RBP analysis pipeline and web platform (PBS Finder) implemented and tested. PBS Finder has been in production for several months; during this time it was thoroughly tested by IBMC experts.

- RNA-Seq analysis pipeline implemented and tested (iRAP deployed and result consolidation tool implemented).
- Integration of both tools could not be accomplished.

## Future Work

#### Conclusions

■ Fully integrate the RNA-Seq analysis pipeline with the web platform (automatic job configuration, result visualization, etc.).

Study the requirements for deploying the platform in large scale, and assess the feasibility of making it available internet-wide.

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