#### Computational translational epigenetics: concept-driven omics analyses

Part 1: The awakening of silent genes in malignancies : a new biomarker discovery strategy

> Part 2: Concept driven omics analyses : EpiMed information system and pipelines



#### Part 1: The awakening of silent genes in malignancies : a new biomarker discovery strategy

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CENTRE DE RECHERCHE UGA - INSERM U 1209 - CNRS UMR 5309

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## **Cell identity**

# **Cell identity**/function is defined by specific gene expression programs



Gene expression programs are controlled by the epigenome = genome signposting system









#### Some gene expression programs are highly **TISSUE-SPECIFIC**



Ex. male germ cells specific genes are **epigenetically "locked"** into a silent state in normal adult somatic cells



Part 1 - biomarker discovery





Genetic and epigenetic alterations







Genetic and epigenetic alterations



Gene repression Tumour suppressors, etc..



Genetic and epigenetic alterations



#### Activation of tissue specific genes





#### Activation of tissue specific genes



#### Activation of tissue specific genes





Activation of tissue specific genes



1 – Very restricted pattern of expression

2 – Unknown to the immune system

Placenta



#### Hypothesis: "out of context" activations of genes occur in all cancer types



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Systematic search for "out of context" (= ectopic) gene expression in cancer

#### 1st step: establishment of a list of « silent » genes



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522 germline- and placenta- restricted genes (TSPS) Part 1 - biomarker discovery



#### 522 germline and placenta « restricted » genes

2nd step: detection of ectopic activation of germline/placenta specific genes in various cancer types



#### 522 germline and placenta « restricted » genes

2nd step: detection of ectopic activation of germline/placenta specific genes in various cancer types



Analysis of approx. 2000 samples from 16 different solid tumours (GSE2109:Expression Project for Oncology (expO) project) Affymetrix microarrays

#### 522 germline and placenta « restricted » genes

2nd step: detection of ectopic activation of germline/placenta specific genes in various cancer types



ON/OFF threshold for binary data

#### 3<sup>rd</sup> step Clinical use of ectopic gene expressions in cancer

#### 3<sup>rd</sup> step Clinical use of ectopic gene expressions in cancer



## **Biomarkers discovery**

New approach to cancer prognosis and treatment





## **Epigenetic origin of ectopic expressions**



Epigenetic abnormalities are responsible for the derepression and aberrant expression of germline/placenta-specific genes in cancer (lung)

=> DNA demethylation of the promoter of germline/placenta genes is associated with ectopic expression



**Evaluation of the clinical value of ectopic expressions as prognosis markers** 



**Evaluation of the clinical value of ectopic expressions as prognosis markers** 

Tumours with ectopic expression

=> Compare survival

Tumours with no expression



**Evaluation of the clinical value of ectopic expressions as prognosis markers** 



=> Gene expression **not** associated with prognosis



**Evaluation of the clinical value of ectopic expressions as prognosis markers** 



Tumours with no expression

Tumours with ectopic expression

=> Gene expression associated with prognosis

« off context » activation of germline restricted genes and prognosis:





**Discriminating power of our ectopic genes classifying system: example (lung)** 

![](_page_38_Figure_2.jpeg)

## **Biomarkers based molecular characterization**

![](_page_39_Figure_1.jpeg)

Gene expression prom

# Biomarkers based molecular characterization P1 P3

Supervised transcriptomic analysis

### Biomarkers based molecular characterization P1 P3

![](_page_41_Figure_1.jpeg)

Up in P3

## **Biomarkers based molecular characterization**

![](_page_42_Figure_1.jpeg)

GeneSet Enrichment Analysis

#### Molecular characterization of highly aggressive tumors

![](_page_42_Figure_4.jpeg)

### Computational translational epigenetics: concept-driven omics analyses

Part 1: The awakening of silent genes in malignancies : a new biomarker discovery strategy

#### Part 2: Concept driven omics analyses : EpiMed information system and pipelines

http://epimed.univ-grenoble-alpes.fr/database/

iab 🤅

EpiMed

![](_page_44_Picture_1.jpeg)

Saadi Khochbin

Dept Signalling and chromatin

![](_page_44_Picture_4.jpeg)

Saadi Khochbin

#### Epigenetics and cell signaling Saadi Khochbin (DRHC, CNRS)

Sophie Rousseaux (*DR2, INSERM*) Anne-Laure Vitte (*IE Bio, UGA*) Ekaterina Flin (*IR bioinfo, UGA*) Florent Chuffart (*IR Bioinfo, INSERM*)

![](_page_44_Picture_8.jpeg)

Sophie Rousseaux

Computational Translational Epigenetics

**EpiMed** 

![](_page_44_Picture_11.jpeg)

Ekaterina Flin

![](_page_44_Picture_13.jpeg)

Florent Chuffard

![](_page_44_Picture_15.jpeg)

Anne-Laure Vitte

![](_page_45_Figure_1.jpeg)

to epigenetics

Florent Chuffart

![](_page_46_Figure_1.jpeg)

![](_page_46_Picture_2.jpeg)

![](_page_46_Picture_3.jpeg)

Computational biology to epigenetics

![](_page_46_Picture_5.jpeg)

![](_page_46_Picture_6.jpeg)

Cancer & chronic diseases biomarkers

IAB 🤅

![](_page_47_Picture_1.jpeg)

EpiMed

![](_page_47_Picture_2.jpeg)

Computational biology to epigenetics

![](_page_47_Picture_4.jpeg)

![](_page_47_Picture_5.jpeg)

Cancer & chronic diseases biomarkers

![](_page_47_Picture_7.jpeg)

![](_page_47_Picture_8.jpeg)

Population Epigenetics

![](_page_48_Picture_0.jpeg)

1- identify tissue-specific genes

2- detection of ectopic expression of tissue specific genes in cancer

- 3- correlate ectopic expressions with clinical outcome
- 4- characterize molecular profile of aggressive tumours

1- identify tissue-specific genes

FniMea

![](_page_49_Picture_2.jpeg)

cross different types of data

2- detection of ectopic expression of tissue specific genes in cancer

- 3- correlate ectopic expressions with clinical outcome
- 4- characterize molecular profile of aggressive tumours

1- identify tissue-specific genes

FniMea

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**ON/OFF** binary datasets

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![](_page_50_Picture_8.jpeg)

![](_page_50_Picture_9.jpeg)

1- identify tissue-specific genes

FniMea

cross different types of data

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**ON/OFF** binary datasets

- 3- correlate ectopic expressions with clinical outcome Correlate with clinical data and survival
- 4- characterize molecular profile of aggressive tumours

![](_page_51_Picture_8.jpeg)

![](_page_51_Picture_9.jpeg)

![](_page_51_Picture_10.jpeg)

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FniMea

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Supervised transcriptomic analysis

![](_page_52_Picture_9.jpeg)

![](_page_52_Picture_10.jpeg)

![](_page_52_Figure_11.jpeg)

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FniMed

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**ON/OFF** binary datasets

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Supervised transcriptomic analysis

5- Understand molecular basis of aggressive phenotype by comparing with other expression profiles (Gene Set Enrichment Analysis), and using available tools to explore biological significance (Gene Ontology terms, pathways...)

![](_page_53_Picture_9.jpeg)

![](_page_53_Picture_10.jpeg)

![](_page_53_Picture_11.jpeg)

![](_page_53_Figure_12.jpeg)

![](_page_53_Figure_13.jpeg)

#### **Specificities of our bioinformatic pipelines**

![](_page_54_Picture_1.jpeg)

cross different types of data

**ON/OFF** binary datasets

Correlate with clinical data and survival

Supervised transcriptomic analysis

**GSEA**, ...

![](_page_54_Picture_7.jpeg)

![](_page_54_Figure_8.jpeg)

![](_page_54_Figure_9.jpeg)

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Increasing amounts of OMICs data

Update and scale-up strategies + Develop new pipelines

Increasing power of analysis + address new questions

![](_page_56_Picture_0.jpeg)

**Issues to be solved** 

- 1- Matching gene and tissues annotations and IDs
- 2- Increasing volumes of data (NGS)
- 3 new types of data (RNAseq, ChIPseq..)
- 4- homogenize clinical annotations and data
- 5- Increase efficiency of pipelines for combined analysis of large and heterogeneous datasets

![](_page_57_Picture_0.jpeg)

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![](_page_57_Picture_8.jpeg)

![](_page_57_Picture_9.jpeg)

**Florent Chuffart** 

![](_page_57_Picture_11.jpeg)

**Ekaterina Flin** 

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![](_page_58_Picture_2.jpeg)

**Ekaterina Flin** 

## Navigate through omics and clinical data

#### 1. Gene/genome annotations

Gene-related annotations are extremely **complex** and frequently modified

#### 2. Clinical data and tissue annotations

Clinical data are **heterogeneous**, especially for tissue annotations which are usually different in different datasets

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![](_page_60_Picture_0.jpeg)

## **EpiMed Information System**

![](_page_60_Picture_3.jpeg)

- Different data types (clinical, omics) from different platforms (Affymetrix, Illumina,...) with different data formats (txt, excel, pdf, raw data)
- Various access to public databases (sql, html, xml, json, web services)
- Scripts and pipelines for data analysis (Python, R)

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![](_page_61_Figure_0.jpeg)

![](_page_62_Figure_0.jpeg)

![](_page_63_Picture_0.jpeg)

#### Dealing with more and more omic data...

Update and implement **pipelines for analysis of omic data** of different origin Improve analysis power and efficiency

Development of new **dedicated pipelines** to answer specific biological questions

=> Explore the epigenomic status/landscape of tissue-specific genes using NGS data

![](_page_63_Picture_5.jpeg)

Florent Chuffart

![](_page_63_Picture_7.jpeg)

EpiMed tools : omic analysis toolbox

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=> Hypothesis : Different epigenetic mechanism are involved in the repression of silent genes

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![](_page_66_Picture_0.jpeg)

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http://epimed.univ-grenoble-alpes.fr/downloads/epimed\_open\_course/

https://epimed.github.io/

https://iab.univ-grenoble-alpes.fr/Plateformes/epigenetique-medicale-et-bioinformatique-epimed