Integrated Genomics and Metabolic Diseases Modeling
UMR 8199 (CNRS / Université de Lille 2 / Institut Pasteur de Lille)

#### Longitudinal Genetic Modelling

Revisiting Associations of SNPs Associated with Blood Fasting Glucose in Normoglycemic Individuals

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Statistical Methods for Post Genomic Data February 11-12, 2016









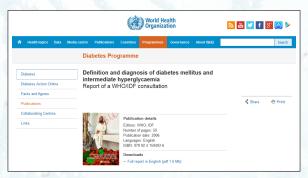
## ${\cal B}$ ackground

- Genome-wide association studies (GWAS) have been extensively used to identify genetic markers.
- Typically, GWAS are based on cross-sectional data.
- The recent use of approaches to account for longitudinal data is driven among other things, by the need to increase statistical power.
- One possible solution is to increase sample size by meta-analysing multiple cohorts [Scott et al., 2012].

#### $\mathcal{D}$ ata

- Metabochip DNA arrays (Illumina) [Voight et al., 2012] assayed in individuals recruited in the French cohort D.E.S.I.R.
   (Données Épidémiologiques sur le Syndrome d'Insulino-Résistance).
- D.E.S.I.R. is a prospective cohort (5,212 individuals) followed up for 9 years (measured at baseline, 3, 6 and 9 years) for many biological traits and several pathologies (e.g. type 2 diabetes, cardiovascular diseases, etc.).

#### ${\mathcal D}$ ata



The focus of this study is on normoglycemic individuals, defined for fasting plasma glucose (FPG) < 7.0 mmol/l [World Health Organization, 2006].

## $\mathcal{D}$ ata

Table: Clinical characteristics of the 5,212 individuals in D.E.S.I.R.

	Men		Women			
	N	mean	sd	N	mean	sd
SEX	2,576	-	-	2,636	-	-
AGE	2,576	46.64	10	2,636	46.9	10.04
BMI	2,531	25.46	3.359	2,584	24.04	4.116
FPG (year = 0)	2,572	5.532	0.8921	2,633	5.188	0.743
FPG (year = 3)	2,234	5.594	0.972	2,264	5.23	0.736
FPG (year = 6)	2,004	5.657	1.033	2,059	5.279	0.7603
FPG (year = 9)	1,953	5.697	1.036	2,024	5.304	0.8011

### ${\mathcal D}$ ata

In this table, we focus on fasting plasma glucose measured at baseline, 3, 6 and 9 years (4 measures).

Table: Missing data rates for the 5,212 individuals in D.E.S.I.R.

	Men	Women
SEX	0.00%	0.00%
AGE	0.00%	0.00%
ВМІ	1.75%	1.97%
FPG (year = 0)	0.155%	0.114%
FPG (year = 3)	13.3%	14.1%
FPG (year = 6)	22.2%	21.9%
FPG (year = 9)	24.2%	23.2%

 $Y_i$  is the measured fasting plasma glucose and  $G_i$  denotes the genotypes coded 0, 1 or 2.

Linear Model (Baseline)

$$m{Y_i} = eta_0 + eta_g m{G_i} + \epsilon_i$$
, where  $\epsilon_i \sim \mathcal{N}(0, \sigma^2)$ 

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$$Y_i = eta_0 + eta_g G_i + \epsilon_i$$
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Linear Model (Average of m measures)

$$ar{Y}_i = eta_0 + eta_g G_i + \epsilon_i$$
, where  $\epsilon_i \sim \mathcal{N}(0, rac{\sigma^2}{m})$ 

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Two-Step

$$\hat{b}_{0i} = \beta_0^* + \beta_q^* G_i + \epsilon_i^*, \text{ where } \epsilon_i^* \sim \mathcal{N}(0, \sigma^{*2})$$

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$$ar{Y}_i = eta_0 + eta_g G_i + \epsilon_i$$
, where  $\epsilon_i \sim \mathcal{N}(0, rac{\sigma^2}{m})$ 

Two-Step

1 
$$Y_{ij} = \beta_0 + b_{0i} + \beta_1 t_{ij} + b_{1i} t_{ij} + \epsilon_{ij}$$
, where  $\epsilon_{ij} \sim \mathcal{N}_m(0, V^{\dagger}_i \equiv Z^{\dagger}_i D^{\dagger} Z^{\dagger}_i' + \sigma^{\dagger^2} I_m)$ 

$$\hat{b}_{0i} = eta_0^* + eta_g^* G_i + \epsilon_i^*$$
, where  $\epsilon_i^* \sim \mathcal{N}(0, \sigma^{*2})$ 

Linear Mixed Model (LMM)

$$Y_{ij} = \beta_0 + b_{0i} + \beta_1 t_{ij} + b_{1i} t_{ij} + \beta_g G_i + \epsilon_{ij}$$
, where  $\epsilon_{ij} \sim \mathcal{N}_m(0, V_i \equiv Z_i D Z_i' + \sigma^2 I_m)$ 

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#### Two-Step

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$$Y_{ij} = \beta_0 + b_{0i} + \beta_1 t_{ij} + b_{1i} t_{ij} + \epsilon_{ij}$$
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2 
$$\hat{b}_{0i} = \beta_0^* + \beta_g^* G_i + \epsilon_i^*$$
, where  $\epsilon_i^* \sim \mathcal{N}(0, \sigma^{*2})$ 

Linear Mixed Model (LMM)

$$Y_{ij} = eta_0 + b_{0i} + eta_1 t_{ij} + b_{1i} t_{ij} + eta_g G_i + \epsilon_{ij}$$
, where  $\epsilon_{ij} \sim \mathcal{N}_m(0, V_i \equiv Z_i D Z_i' + \sigma^2 I_m)$ 

Generalised Estimating Equations (GEE)

$$\mathbb{E}(Y_i) = \beta_0 + \beta_1 t_{ii} + \beta_q G_i$$
 and  $\mathbb{V}(Y_i) = V_i$  (Compound Symmetry).

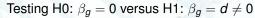
## $\mathcal{T}$ op associated SNPs

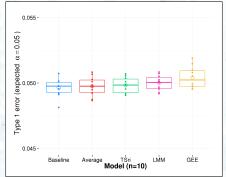
SNP	Chromosome	Position (hg18)	Gene	Minor Allele	MAF	P value
rs560887	2	169,471,394	G6PC2	Т	0.30	1.5e-15
rs2908289	7	44,190,467	GCK	Α	0.19	4.4e-05
rs16913693	9	110,720,180	IKBKAP	G	0.02	4.1e-04
rs2191349	7	15,030,834	DGKB/TMTM195	G	0.44	1.1e-03
rs6072275	20	39,177,319	TOP1	Α	0.13	4.7e-03
rs340874	1	212,225,879	PROX1	T	0.45	7.8e-03
rs3783347	14	99,909,014	WARS	T	0.20	1.3e-02
rs3829109	9	138,376,587	DNLZ	Α	0.26	1.4e-02
rs11607883	11	45,796,285	CRY2	G	0.46	3.0e-02
rs2302593	19	50,888,474	GIPR	G	0.48	3.7e-02

SNPs selected based on Yaghootkar and Frayling [2013] and Vaxillaire et al. [2014].

## ${\cal R}$ esults: Type 1 error

Type 1 error was computed for 10 SNPs previously shown to be significantly associated with fasting plasma glucose in Vaxillaire et al. [2014].





100,000 genotypes permutations were performed for each selected SNP.

## ${\cal R}$ esults: Statistical power (post-hoc)

SNP	Chromosome	Position (hg18)	Gene	Minor Allele	MAF	P value
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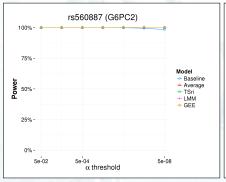
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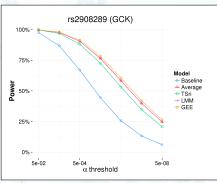
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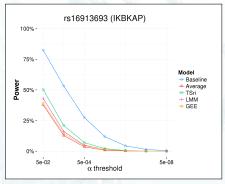
## ${\cal R}$ esults: Statistical power (post-hoc)

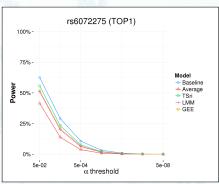




100,000 resamplings were performed for each selected SNP.

## ${\cal R}$ esults: Statistical power (post-hoc)





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# ${\cal N}$ on-Centrality Parameter & statistical power

Closed-form formulas for testing SNP association under the Cross-Sectional (CS), the Random Intercept (RI) and the Random Intercept and Slope model (RIS)

$$NCP_{CS} = nd^2\left(\frac{2p(1-p)}{\sigma^2}\right)$$
 (1)

$$NCP_{RI} = NCP_{CS} \left( \frac{m\sigma^2}{\sigma^2 + m\sigma_{b0}^2} \right)$$
 (2)

$$NCP_{RIS} = NCP_{RI}U$$
 (3)

with 
$$U = \frac{(\sigma^2 + \sigma_{b1}^2 \sum_{j=1}^m (t_j - \bar{t})^2)(\sigma^2 + m\sigma_{b0}^2)}{(\sigma^2 + \sigma_{b1}^2 \sum_{j=1}^m (t_j - \bar{t})^2)(\sigma^2 + m\sigma_{b0}^2) - m\rho^2 \sigma_{b0}^2 \sigma_{b1}^2 \sum_{j=1}^m (t_j - \bar{t})^2} \ge 1$$
 (4)

## ${\cal N}$ on-Centrality Parameter & statistical power

This implies that  $NCP_{RIS} \ge NCP_{RI}$  but no guarantee that  $NCP_{RIS} > NCP_{CS}$ 

SNP	Gene	$NCP_{CS}$	$NCP_{RIS}$ $(NCP_{RI})$
rs560887	G6PC2	63.33	93.08 (92.69)
rs2908289	GCK	17.02	26.37 (26.26)
rs16913693	IKBKAP	12.75	6.55 (6.53)
rs6072275	TOP1	7.78	6.78 (6.76)

## ${\cal N}$ on-Centrality Parameter & statistical power

This implies that  $NCP_{RIS} \ge NCP_{RI}$  but **no guarantee** that  $NCP_{RIS} > NCP_{CS}$ 

SNP	Gene	$NCP_{CS}$		NCP <sub>RIS</sub> (NCP <sub>RI</sub> )
rs560887	G6PC2	63.33	<	93.08 (92.69)
rs2908289	GCK	17.02	<	26.37 (26.26)
rs16913693	IKBKAP	12.75	>	6.55 (6.53)
rs6072275	TOP1	7.78	>	6.78 (6.76)

#### ${\mathcal F}$ uture research

 Characterise parameter space for which NCP<sub>RIS</sub> is indeed higher than NCP<sub>CS</sub>;

- Derive formula for the non-centrality parameter when testing SNP\*time interaction effect;
- Check results consistency according to missing data distribution (MCAR, MAR and MNAR).

## ${\cal A}$ cknowledgements

# Thank you for your attention!

#### **CNRS UMR 8199**

Ghislain Rocheleau Loïc Yengo Philippe Froguel

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