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### Loci underlying variation in nematode resistance in three European sheep populations: a joint-analysis



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- Gastrointestinal nematode infections large impact on the sheep industry:
  - Anthelmintic treatment
  - Production losses





e.g. €100 million/year in UK



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- Anthelmintic resistance has developed fast in many nematode populations:
  - Need for new control measures



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- Nematode resistance is a complex trait:
  - Large number of physiological pathways involved
  - Indicator traits (i.e. Faecal Egg Count) are time specific





- Selection for increased resistance to nematodes has often been suggested
- Nematode resistance is a complex trait:
  - Large number of physiological pathways involved
  - Indicator traits (i.e. Faecal Egg Count) are time specific
- Advantageous to select directly for resistance:
  - Several QTL studies addressed nematode resistance



- Little overall consensus has emerged from these studies in terms of resistance loci:
  - Apparent genetic complexity of the trait
  - Variety of sheep breeds, nematode species and experimental approaches
- Do common regions exist? → meta (or joint) analysis
  - Tool for aggregating information from multiple independent studies





SEVENTH FRAMEWORK

# Aim

# Identify genomic regions underlying FEC variation in a joint analysis of three European sheep populations









#### • Data

- Average animal effect for Strongyles FEC on 4123 individuals from the three populations
  - 752 Scottish Blackface (SBF) lambs
  - 2371 Sarda x Lacaune backcross (SAR) ewes
  - 1000 Martinik Black-Belly x Romane backcross (MBR) lambs





#### Data

- Different Strongyles species and challenges:
  - natural (mixed species) challenge at pasture for SBF
    - mainly Teladorsagia
  - natural (mixed species) challenge at pasture for SAR
    - changes through the year
  - artificial challenge with *Haemonchus* for MBR



#### Data

- Fixed effects specific for each population
- Animals genotyped with the 50k SNPchip
- QC specific for each population:
  - 38,991 SNPs in common after QC
- SNP positions from Sheep Genome browser v2.0





- QTL found in previous population-specific analyses:
  - SBF:
    - Chr 3 & 6
  - SAR:
    - Chr 7, 12 & 20
  - **MBR**:
    - Chr 5, 12 & 13



- Regional Heritability Mapping (RHM)
  - Variance component approach
  - Fit joint effects of all loci within a genomic region
  - Each chromosome is divided into windows of a pre-defined number of SNPs:







![](_page_13_Figure_2.jpeg)

$$y = Xb + Za + Zw + e$$

overall genetic effect

regional combined genetic effect

$$h^{2} = \frac{\left(\sigma_{a}^{2} + \sigma_{w}^{2}\right)}{\left(\sigma_{a}^{2} + \sigma_{w}^{2} + \sigma_{e}^{2}\right)}$$

$$n_w^2 = \frac{\sigma_w^2}{\left(\sigma_a^2 + \sigma_w^2 + \sigma_e^2\right)}$$

#### Regional h<sup>2</sup>

![](_page_13_Picture_9.jpeg)

Total h<sup>2</sup>

![](_page_13_Picture_11.jpeg)

### • Problem 1:

Unrelated populations and therefore IBS relationships uninformative

### • Problem 2:

 – few sire families and therefore long chromosome segments inherited intact → long stretches of LD

![](_page_14_Picture_6.jpeg)

![](_page_14_Picture_7.jpeg)

# **Material and Methods - Solutions**

### Problem 1

- Unrelated populations and therefore IBS relationships uninformative
- Genomic relationship matrix (G) set to block diagonal → covariance between populations = 0

![](_page_15_Picture_5.jpeg)

![](_page_15_Picture_6.jpeg)

• Problem 2 Region of interest

![](_page_16_Picture_3.jpeg)

![](_page_16_Picture_4.jpeg)

![](_page_17_Figure_2.jpeg)

![](_page_18_Figure_2.jpeg)

![](_page_18_Picture_3.jpeg)

![](_page_18_Picture_4.jpeg)

# Material and Methods - Solutions

- Problem 1
- Problem 2
  - few sire families and therefore long chromosome segments inherited intact → long stretches of LD
- Two different G matrices:
  - whole  $\rightarrow$  using all SNPs across the genome
  - n-1 → created separately for each chromosome excluding the chromosome being interrogated

![](_page_19_Picture_8.jpeg)

![](_page_19_Picture_9.jpeg)

- Model tested by Likelihood Ratio Test (LRT):
  compared to LogL of model with no QTL
- Correction for multiple testing required:
  - threshold for genome-wide (p<0.05) significance 13.38</p>
  - threshold for suggestive significance 9.11

![](_page_20_Picture_6.jpeg)

![](_page_20_Picture_7.jpeg)

### **Results**

#### Plot of all genome using the whole genomic relationship matrix

![](_page_21_Figure_3.jpeg)

OAR	window	LRT	h <sup>2</sup> <sub>w</sub>
20	8	13.78	0.02
20	9	16.50	0.02
20	10	13.88	0.02
4	2	10.24	0.01

![](_page_21_Picture_5.jpeg)

![](_page_21_Picture_6.jpeg)

### **Results**

#### Plots of all genome using the n-1 genomic relationship matrix

![](_page_22_Figure_3.jpeg)

OAR	window	LRT	h <sup>2</sup> w
20	8	21.28	0.02
20	9	23.52	0.02
20	10	20.74	0.02
4	2	14.40	0.02
19	10	13.74	0.02

![](_page_22_Picture_5.jpeg)

![](_page_22_Picture_6.jpeg)

### **Results**

#### Plots of all genome using the n-1 genomic relationship matrix

![](_page_23_Figure_3.jpeg)

![](_page_23_Picture_4.jpeg)

![](_page_23_Picture_5.jpeg)

# **Considerations on data structure**

- The n-1 G partially overcomes problem of long stretches of LD
- RHM can detect different types of QTL architecture from LA or LD methods:
  - does not require large differences between contrasting loci in same population
- MHC was most significant region:
  - Region characterised by extreme complexity

![](_page_24_Picture_7.jpeg)

# Conclusions

- Despite heterogeneity of data, joint-analysis allowed identification of common regions
- Using a whole (block diagonal) G, some QTL significant in individual datasets were lost

 Using the n-1 G, the QTL from individual studies reappeared, and those already found become more significant

![](_page_25_Picture_5.jpeg)

![](_page_25_Picture_6.jpeg)

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![](_page_26_Picture_1.jpeg)

![](_page_26_Picture_2.jpeg)

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![](_page_26_Picture_5.jpeg)

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