# ibdreg: An R package for Genetic Linkage with Covariates

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 Intro to Linkage (brief) Why do linkage with covariates? Methodology in ibdreg Demo of ibdreg on prostate cancer data

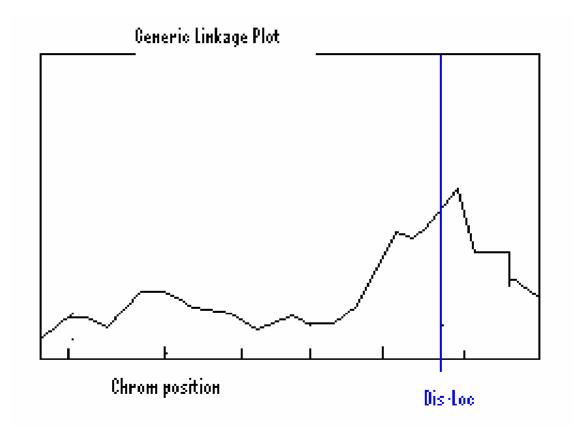
#### Linkage Intro: Genetic Analysis Types

- Segregation Analysis Use family data to determine the extent to which a disease is genetic
- Linkage Analysis Use family data to locate a chromosome segment that is inherited jointly with a disease locus
- Association Analysis Use unrelated data to find a genetic effect of a particular DNA marker(s) on a disease

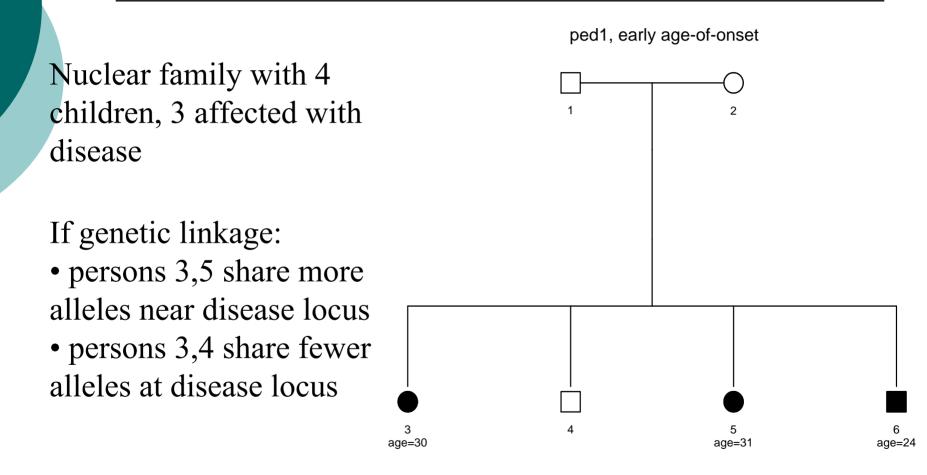
# Intro to Linkage: Recombination

- Mendel proposed genes (on peas) are inherited independently of one another
   Not quite right -- genes / markers which
- are "close" together show associations
- During meiosis, corresponding DNA segments can recombine in any location
- Recombination is somewhat constant over the whole genome
- In general, "close" markers have fewer recombinations

# Intro to Linkage, basic plot



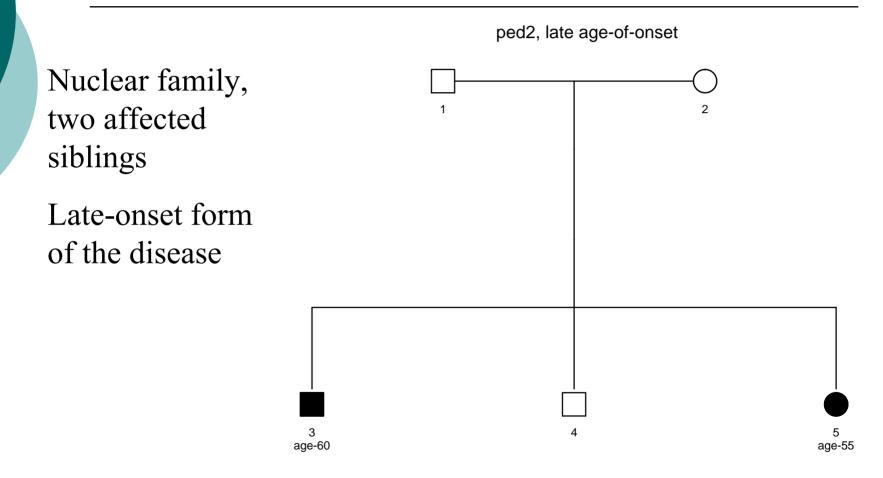
# Linkage with Covariates: illustrated



#### Linkage with Covariates: IBD Sharing

- Identity-By-Descent (IBD): alleles that have been inherited from the same ancestor
- Estimate probability of sharing 0, 1, and 2 alleles IBD: f<sub>0</sub>, f<sub>1</sub>, f<sub>2</sub>
- <sup>o</sup> Use Merlin, Genehunter, etc.
- Estimated IBD sharing for a relative pair:  $s_r = 2f_{r,2} + 1f_{r,1}$

### Linkage with Covariates: illustrated



# Linkage with Covariates

- Ped1 and Ped2 could be two forms of disease: early and late onset
- Mixture of gene and environment effects, i.e. disease heterogeneity
- Assume the early onset is genetically linked at locus L
- IBD sharing at L is:

ped1: greater than null

ped2: somewhere near null

# Linkage with Covariates

- The more alike the covariates, the more allele sharing is expected
- Consider a function of the age of onset for affected siblings:
  - 1. age1 + age2
  - 2. age1 \* age2
  - 3. (|age1-60| \* |age2-60|)<sup>2</sup>
- Define any function that applies to the disease and covariate impact

#### Linkage with Covariates

- Intuitively, affected relative pairs with similar covariates share more alleles than the null
- Linear relationship for IBD sharing and pair-specific covariates (X):

$$m_r = m_r^0 + c_r X\beta = X_r^*\beta$$

m<sub>r</sub>, m<sub>r</sub><sup>o</sup>: Expected s<sub>r</sub> under linkage, no linkage c<sub>r</sub> = scaling factor for the relationship types  $X_r^* = c_r[1 | X]$ 

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# ibdreg method

Quasi-likelihood score function:

$$U = \sum_{i=1}^{n} D_i^T V_i^{-1} (S_i - M_i)$$

where

- i = 1..n pedigrees
- $S_i$ ,  $M_i$ : vectors of estimated and expected ( $H_o$ ) allele sharing
- $D_i$ : derivative of  $M_i$  with respect to  $\beta$
- $V_i$ : covariance matrix of the  $S_i$  vector

# ibdreg method

Develop a chi-square test:  $T = U^T V_u^{-1} U$ 

where 
$$V_{u} = \sum_{i=1}^{n} X_{i}^{*T} V_{0,i}^{-1} X_{i}^{*}$$

With q degrees of freedom: q=1 for intercept-only (linkage) q=2 for intercept with 1 covariate etc.

# ibdreg method

Available tests (degrees of freedom)

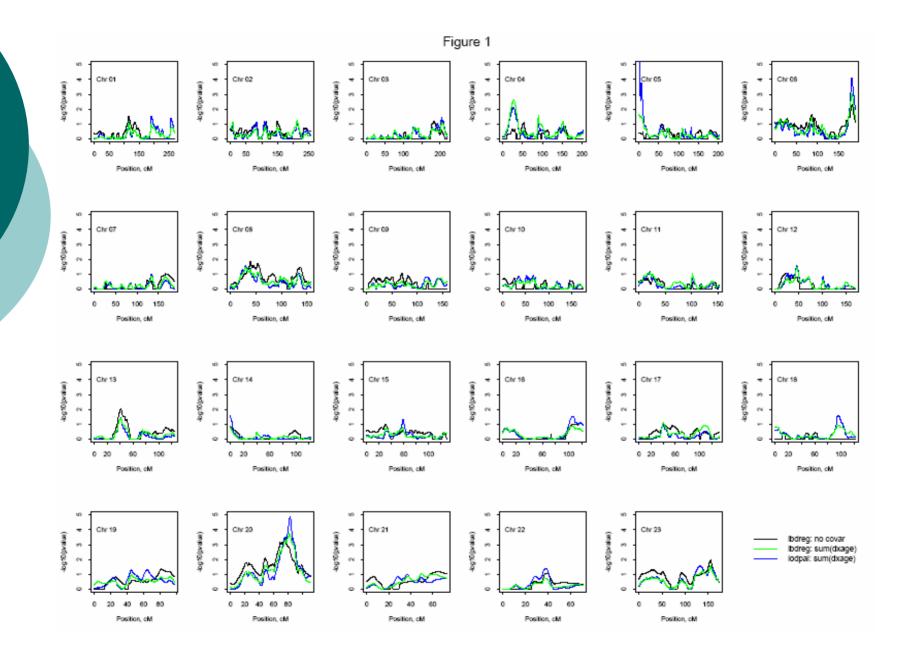
- 1. Linkage without covariates (q=1)
- 2. Linkage with p covariates (q=1+p)
- 3. Effect of p covariates on IBD sharing, adjusting for linkage (p)

#### ibdreg applied

- Data: sample of pedigrees ascertained at Mayo Clinic (SE MN) for having multiple cases of prostate cancer
- 159 pedigrees (max 21 people)
- 495 affected relative pairs (ARPs)
- Covariate: age-of-onset (dxAge)

## ibdreg applied

- LODPAL, of S.A.G.E., approximates a pseudo-likelihood of IBD sharing probabilities by a trinomial logistic regression model (blue line)
- Compare to ibdreg (green line), both using sum(dxAge) as a covariate
- Also include ibdreg, linkage-only (black line)
- Same c scale (genetic effect between dominant and recessive)



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# ibdreg applied

For figure above, note a few things:

- LODPAL and ibdreg generally close
  Differences:
  - 1. Narrow peaks (ch6, ch20)
  - 2. Ends of chromosomes (ch5)
- Both may be attributable to instability of maximizing pseudolikelihood of LODPAL

# ibdreg applied: R code

```
# load the library
R> library(ibdreg)
```

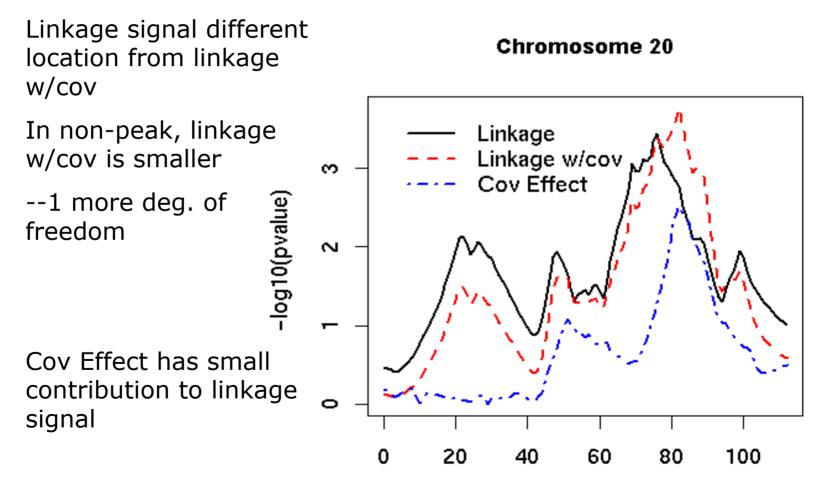
# load covariate data, containing ped id and person id to match file
R> cov.data <- read.table("cov.data.csv", sep=",")</pre>

## ibdreg applied: R code

```
# define a function for the sum of covariates
R> pairSum <- function(cov1,cov2) {cov1+cov2}</pre>
```

# run ibdreg for AA relatives, # with covariate sum(dxage), minimax c scaling R> sum.dxage.AA <ibdreg(formula=~pairSum(dxage), status.method="AA", c.scale="minimax", status=pcstat, ped.id=ped.id, person.id=person.id, data=cov.data, ibd.dat=ibd.dat.obj, ibd.var=ibd.var.obj)

#### ibdreg applied: chromosome 20



#### ibdreg applied: all ARPs

- Disregarding covariates, we can test linkage on all 495 relative pairs in the example, which contain sub-groups
  - AA: Affected-Affected (429)
  - UU: Unaffected-Unaffected (7)
  - AU: Affected-Unaffected (59)

#### • Sharing under linkage is expected as

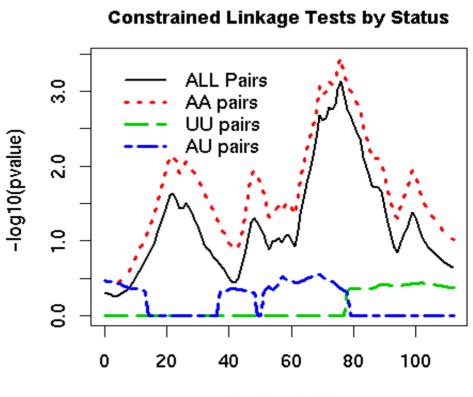
- AA: More than null
- UU: More than null, but less than AA
- UU: Less than null

#### Ibdreg applied, all ARPs

Linkage tests for all relative pairs

Test for linkage in direction expected under linkage: AA-more UU-more AU-less

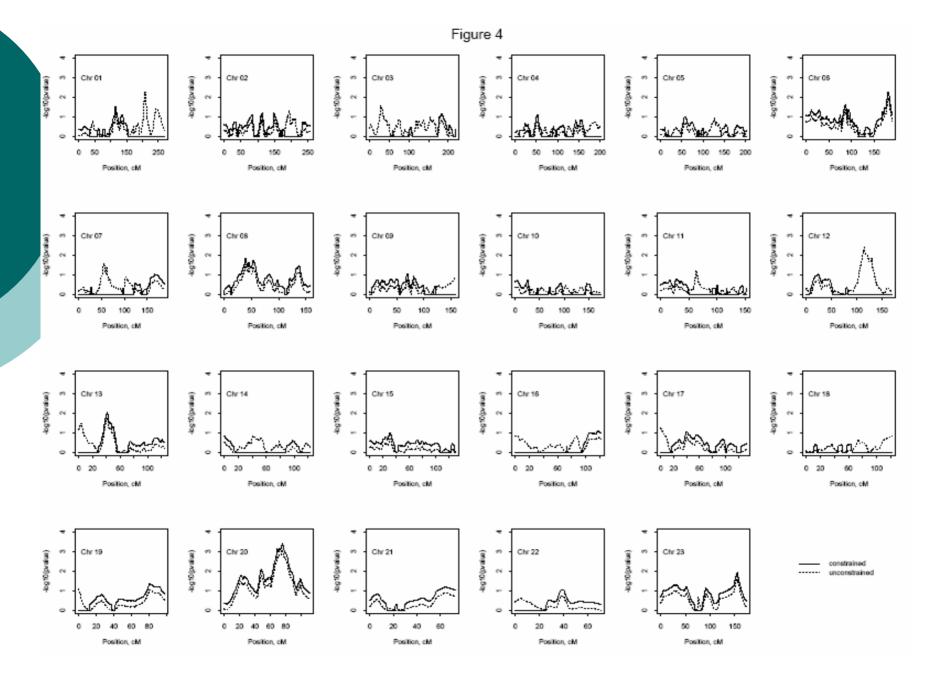
Overall linkage is driven by AA pairs



Position(cM)

# ibdreg applied: unexpected sharing

- IBD allele sharing can be invalid by:
  - 1. Mis-specify marker allele freq when parents missing
  - 2. Incorrect pedigree relationships
  - 3. Unaccounted inbreeding
- The  $\chi^2$  test for linkage given as both unconstrained and constrained to favor linkage (1-sided, smaller p-value)
- If p<sub>unconstrained</sub> < p<sub>constrained</sub>, indicates unexpected IBD sharing (chroms 1, 7, 12)



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#### ibdreg, concluded

- Allows multiple tests for linkage
- $_{\rm 0}$  Provides options for specifying covariates and scaling (c\_r)
- Scaling avoids fitting new regression line for each relationship type
- Linear regression is easy to fit, and easy to apply weights to account for dependence of variance on covariates
- Returned S3 object with print and plot methods
- Perl scripts supplied to work with Merlin.

# Thank You

#### For more details and references see:

Schaid DJ, Sinnwell JP, Thibodeau SN. Testing genetic linkage with relative pairs and covariates by quasilikelihood score statistics. Hum Hered. 2007;64(4):220-33. Epub 2007 Jun 12.