

Model Fitting

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Seattle, June 2008

Regression commands

Two of the most important R commands;

• lm(): fits Linear Models

• glm(): fits **G**eneralized **L**inear **M**odels

(If you've used SAS, its glm is **not** the same as R's)

'Linear Regression' and 'Logistic Regression' are special cases.

The help files are huge (and generic) — how are lm(), glm() used in genetics?

For a continuous outcome,

$$lm(outcome \sim genetic.predictor, [...])$$

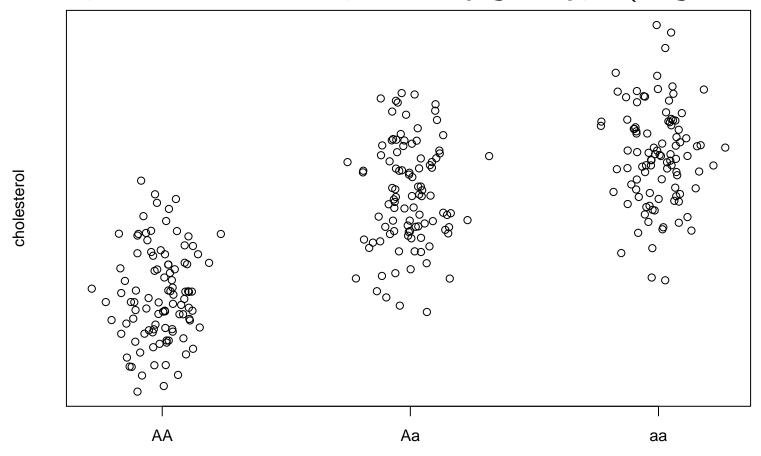
estimates the association between outcome and predictor

The **optional** arguments [...] might be

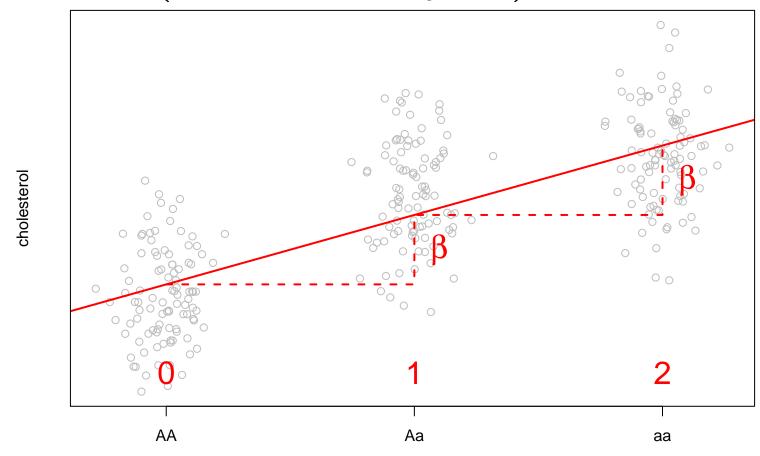
- data=my.data your dataset
- subset=race==CEPH use partial data
- weights for advanced analyses

| Model Description | predictor | Common name |
|-----------------------------|-------------------------|--------------|
| Number of minor alleles | (g=='Aa') + 2*(g=='aa') | Additive |
| | Or as.numeric(g) | |
| Presence of minor allele | (g=='Aa') (g=='aa') | Dominant |
| Homozygous for minor allele | g=='aa' | Recessive |
| Distinct effects | factor(g) | 2 parameter, |
| for hetero/homozygous | | or "2 df" |

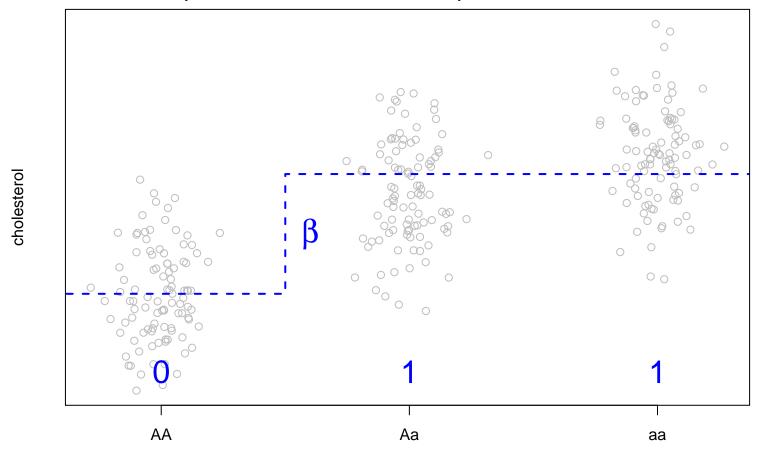
Some data; cholesterol levels plotted by genotype (single SNP)



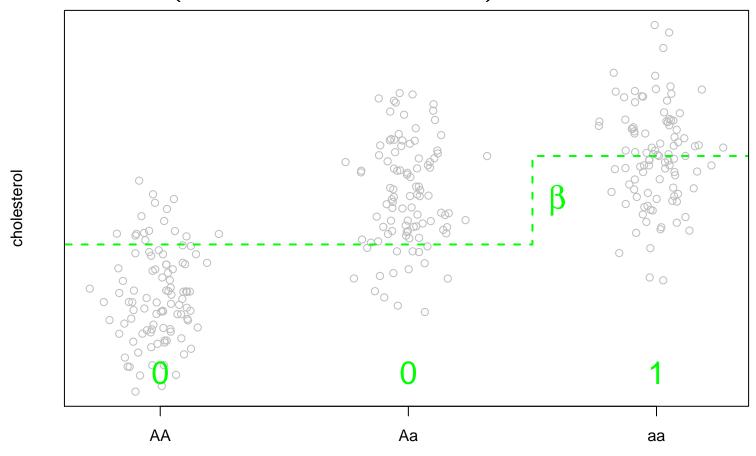
Additive model (the most commonly used)



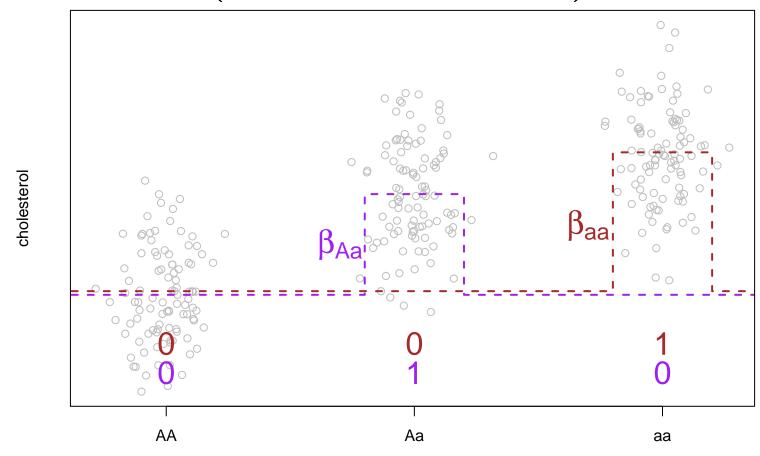
Dominant model (best fit to this data)



Recessive model (least stable for rare aa)



2 parameter model (robust but can be overkill)



lm(): Estimates, Intervals, p-values

lm() produces point estimates for your model;

also available via my.lm\$coefficients.

The coefficients in the output tell you the **additive increase** in outcome associated with a **one-unit** difference in the genetic predictor.

The coefficient for n.minor is in units of cholesterol

lm(): Estimates, Intervals, p-values

You will also want confidence intervals;

Remember to **round these numbers** to an appropriate number of significant figures! (2 or 3 is usually enough)

We are **seldom** interested in the Intercept

lm(): Estimates, Intervals, p-values

Two-sided **p-values** are also available;

In this data, we have **strong evidence** of an **additive effect** of the minor allele on cholesterol

summary(my.lm) gives many other details — ignore for now

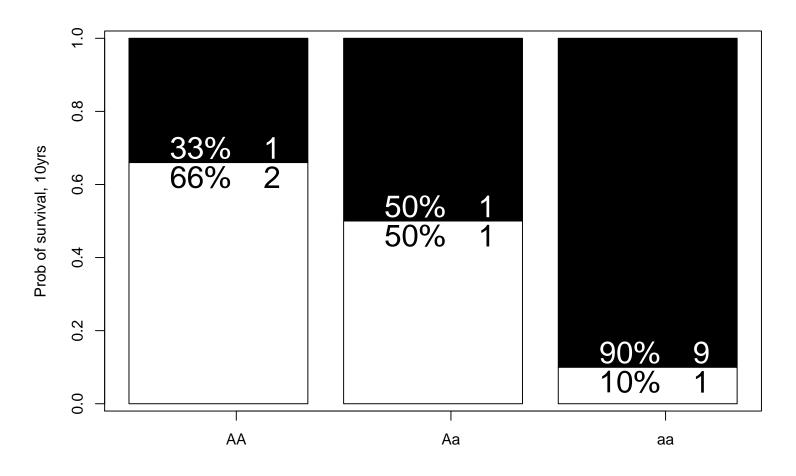
Confidence intervals are just Estimate $\pm 2 \times \text{Std.Error}$

Logistic regression is the 'default' analysis for binary outcomes

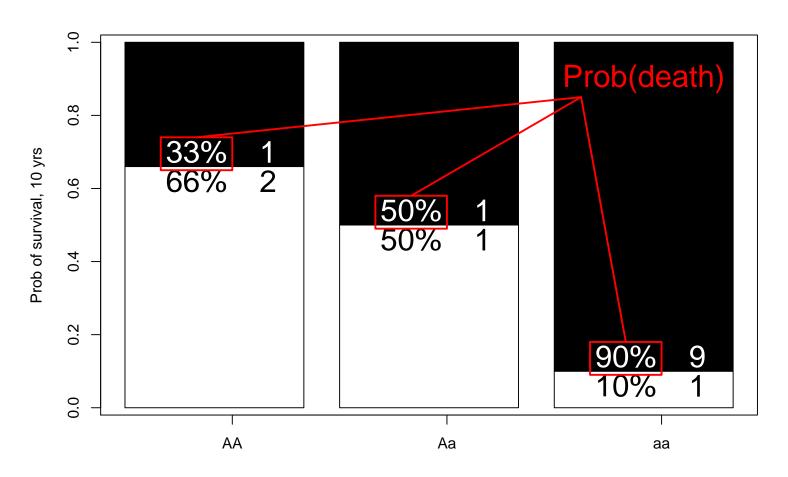
| Outcome | Туре | Regression | Scale |
|----------------------------|------------|------------|-----------------------|
| Cholesterol Blood Pressure | Continuous | Linear | Difference in Outcome |
| BMI | Continuous | Lillear | Difference in Outcome |
| Death Stroke BMI>30 | Binary | Logistic | Ratio of odds |

What are **odds**? Really just **probability**...

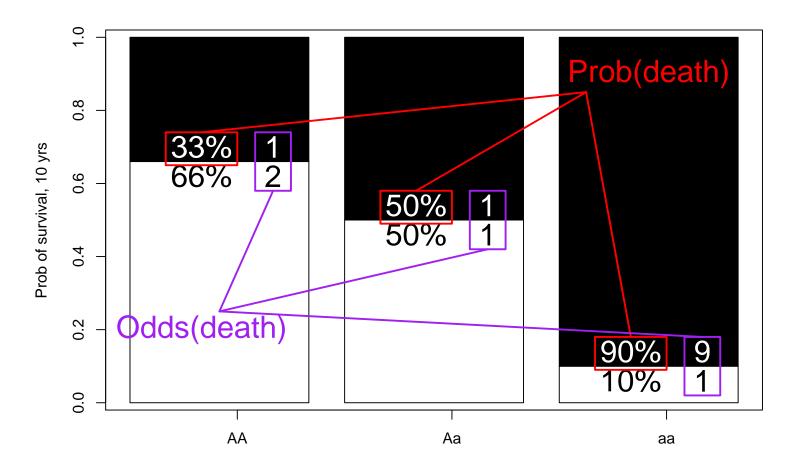
Odds are a [gambling-friendly] measure of chance;



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Odds are a [gambling-friendly] measure of chance;



– so what are odds ratios?

Using the data from the bar charts;

These are \log odds ratio estimates; to transform to OR, use $e^{0.6931}=2, e^{2.8904}=18$

They are given **relative to the baseline group** — 'AA' in this case

Don't forget the family=binomial argument!

Confidence intervals, p-values as with lm(), for the log odds ratios;

Use exp() to get odds ratio estimates, intervals; p-values are scale-independent