

A more meaningful question is whether smokers are different from non-smokers when holding other variables constant, such as age. To address this question, you need to add age into the model.

It might be natural to consider each age — 35, 36, 37, and so on — as a separate group, but you won't get very many members of each group. And, likely, the data for 35 year-olds has quite a lot to say about 36 year-olds, so it doesn't make sense to treat them as completely separate groups.

You can use the `cut()` function to divide up a quantitative variable into groups. You get to specify the breaks between groups. Using `transform()`, you can add the new variable to an existing data frame.

```
> w = transform(w, ageGroups=cut(age,breaks=c(0,30,40,53,64,75,100)))
> mean( outcome=="Alive" ~ ageGroups, data=w )
```

(0,30]	(30,40]	(40,53]	(53,64]	(64,75]	(75,100]
0.979	0.948	0.832	0.625	0.201	0.000

```
> mean( outcome=="Alive" ~ smoker + ageGroups, data=w )
```

No. (0,30]	Yes. (0,30]	No. (30,40]	Yes. (30,40]	No. (40,53]	Yes. (40,53]
0.982	0.976	0.955	0.941	0.876	0.802
No. (53,64]	Yes. (53,64]	No. (64,75]	Yes. (64,75]	No. (75,100]	Yes. (75,100]
0.669	0.581	0.214	0.158	0.000	0.000

The mean has been calculated group-by-group. This is a very widely used technique, but there is a better approach that will be introduced in later chapters: use quantitative variables directly without dividing them into groups.

4.6.1 Model Values and Residuals

A group-wise model tells you the model value for each group. There is additional information that you will want to generate about models. Two fundamental aspects of a model are the **fitted model values** for each case and the **residual** for each case. To make it easy to do these calculations, R has a set of modeling functions that keep track of the data used in creating the model. Later chapters will introduce the `lm()` function — l for "linear", m for "model" — that is central to statistical modeling. To create a model based on groupwise means, use the `mm()` function — the first m for means, the second m for "model." `mm()` does the same sorts of calculations as `mean()`, but packages up its results in a different way:

```
> kids = fetchData("kidsfeet.csv")
> mod = mm( width ~ sex, data=kids )
> mod
```

Groupwise Model

Coefficients:

B	G
9.19	8.78