Odds Ratios

- The ratio of the probability of choosing one outcome category over the probability of choosing the baseline category is often referred as relative risk
- It is also sometimes referred as odds.

- ► The relative risk is the (right-hand side) linear equation exponentiated, leading to the fact that the exponentiated regression coefficients are relative risk ratios for a unit change in the predictor variable.
- We can exponentiate the coefficients from our model to see these odds ratios (next slide).

Extract the coefficients from the model then and exponentiate

```
exp(coef(test))

(Intercept) sesmiddle seshigh write general 17.33 0.5867 0.3126 0.9437 vocation 184.61 1.3383 0.3743 0.8926
```

- ► The relative risk ratio for a one-unit increase in the variable **write** is 0.9437 for being in general program vs. academic program.
- ► The relative risk ratio switching from ses = 1 to 3 is 0.3126 for being in general program vs. academic program.

- You can also use predicted probabilities to help you understand the model.
- You can calculate predicted probabilities for each of our outcome levels using the fitted function.
- We can start by generating the predicted probabilities for the observations in our dataset and viewing the first few rows

```
head(pp <- fitted(test))</pre>
```

```
academic general vocation
1 0.1483 0.3382 0.5135
2 0.1202 0.1806 0.6992
3 0.4187 0.2368 0.3445
4 0.1727 0.3508 0.4765
5 0.1001 0.1689 0.7309
6 0.3534 0.2378 0.4088
```

Prediction

- Suppose we want to examine the changes in predicted probability associated with one of our two variables, we can create small datasets varying one variable while holding the other constant.
- We will first do this holding write at its mean and examining the predicted probabilities for each level of ses.
- ▶ (i.e. Three Cases to predict for)

0.4397 0.3582 0.2021

2 0.4777 0.2283 0.2939 3 0.7009 0.1785 0.1206

Another way to understand the model using the predicted probabilities is to look at the averaged predicted probabilities for different values of the continuous predictor variable write within each level of ses.

Store the predicted probabilities for each value of ses and write

```
dwrite <- data.frame(
  ses = rep(c("low", "middle", "high"), each = 41),
  write = rep(c(30:70), 3))

pp.write <- cbind(dwrite,
      predict(test, newdata = dwrite,
      type = "probs", se = TRUE))</pre>
```

Calculate the mean probabilities within each level of ses

```
by(pp.write[, 3:5], pp.write$ses, colMeans)
```

```
pp.write$ses: high
academic general vocation
 0.6164 0.1808 0.2028
pp.write$ses: low
academic general vocation
 0.3973 0.3278 0.2749
pp.write$ses: middle
academic general vocation
 0.4256 0.2011 0.3733
```

- ▶ A couple of plots can convey a good deal amount of information.
- Using the predictions we generated for the pp.write object above, we can plot the predicted probabilities against the writing score by the level of ses for different levels of the outcome variable.

```
remark: melt data set to long for ggplot2
lpp <- melt(pp.write, id.vars = c("ses", "write"),</pre>
           value.name = "probability")
head(lpp) # view first few rows
  ses write variable probability
 1 low
         30 academic 0.09844
2 low 31 academic 0.10717
3 low 32 academic 0.11650
4 low 33 academic 0.12646
5 low 34 academic 0.13705
6 low 35 academic 0.14828
```

each level of ses facetted by program type

ggplot(lpp aes(x = write y = probability o

plot predicted probabilities across write values for