

Feedback — Quiz 4 **Please Note: No Grace Period** [Help Center](#)

You submitted this quiz on **Thu 7 May 2015 11:33 PM PDT**. You got a score of **6.00** out of **6.00**. However, you will not get credit for it, since it was submitted past the deadline.

Question 1

Consider the space shuttle data `?shuttle` in the `MASS` library. Consider modeling the use of the autolander as the outcome (variable name `use`). Fit a logistic regression model with autolander (variable `auto`) use (labeled as "auto" 1) versus not (0) as predicted by wind sign (variable `wind`). Give the estimated odds ratio for autolander use comparing head winds, labeled as "head" in the variable `headwind` (numerator) to tail winds (denominator).

Your Answer	Score	Explanation
<input type="radio"/> -0.031		
<input checked="" type="radio"/> 0.969	1.00	
<input type="radio"/> 0.031		
<input type="radio"/> 1.327		
Total	1.00 / 1.00	

Question Explanation

```
library(MASS)
data(shuttle)
## Make our own variables just for illustration
shuttle$auto <- 1 * (shuttle$use == "auto")
shuttle$headwind <- 1 * (shuttle$wind == "head")
fit <- glm(auto ~ headwind, data = shuttle, family = binomial)
exp(coef(fit))
```

```
## (Intercept)    headwind
##           1.3273      0.9687
```

```
## Another way without redifing variables  
fit <- glm(relevel(use, "noauto") ~ relevel(wind, "tail"), data = shuttle, family = binomial)  
exp(coef(fit))
```

```
## (Intercept) relevel(wind, "tail")head  
## 1.3273 0.9687
```

Question 2

Consider the previous problem. Give the estimated odds ratio for autolander use comparing head winds (numerator) to tail winds (denominator) adjusting for wind strength from the variable magn.

Your Answer	Score	Explanation
<input type="radio"/> 0.684		
<input type="radio"/> 1.00		
<input checked="" type="radio"/> 0.969	✓ 1.00	
<input type="radio"/> 1.485		
Total	1.00 / 1.00	

Question Explanation

The estimate doesn't change with the inclusion of wind strength

```
shuttle$auto <- 1 * (shuttle$use == "auto")  
shuttle$headwind <- 1 * (shuttle$wind == "head")  
fit <- glm(auto ~ headwind + magn, data = shuttle, family = binomial)  
exp(coef(fit))
```

```
## (Intercept) headwind magnMedium magnOut magnStrong  
## 1.4852 0.9685 1.0000 0.6842 0.9376
```

```
## Another way without redifing variables  
fit <- glm(relevel(use, "noauto") ~ relevel(wind, "tail") + magn, data = shuttle,  
family = binomial)  
exp(coef(fit))
```

```
## (Intercept) relevel(wind, "tail")head
```

```
##          1.4852      0.9685
## magnMedium      magnOut
##          1.0000      0.6842
## magnStrong
##          0.9376
```

Question 3

If you fit a logistic regression model to a binary variable, for example use of the autolander, then fit a logistic regression model for one minus the outcome (not using the autolander) what happens to the coefficients?

Your Answer	Score	Explanation
<input type="radio"/> The intercept changes sign, but the other coefficients don't.		
<input checked="" type="radio"/> The coefficients reverse their signs.	✓ 1.00	
<input type="radio"/> The coefficients get inverted (one over their previous value).		
<input type="radio"/> The coefficients change in a non-linear fashion.		
Total	1.00 / 1.00	

Question Explanation

Remember that the coefficients are on the log scale. So changing the sign changes the numerator and denominator for the exponent.

Question 4

Consider the insect spray data [InsectSprays](#). Fit a Poisson model using spray as a factor level. Report the estimated relative rate comparing spray A (numerator) to spray B (denominator).

Your Answer	Score	Explanation
<input type="radio"/> 0.136		
<input checked="" type="radio"/> 0.9457	✓ 1.00	
<input type="radio"/> -0.056		

0.321

Total 1.00 / 1.00

Question Explanation

```
fit <- glm(count ~ relevel(spray, "B"), data = InsectSprays, family = poisson)
exp(coef(fit))[2]
```

```
## relevel(spray, "B")A
##          0.9457
```

Question 5

Consider a Poisson glm with an offset, t . So, for example, a model of the form `glm(count ~ x + offset(t), family = poisson)` where `x` is a factor variable comparing a treatment (1) to a control (0) and `t` is the natural log of a monitoring time. What is impact of the coefficient for `x` if we fit the model `glm(count ~ x + offset(t2), family = poisson)` where `t2 <- log(10) + t`? In other words, what happens to the coefficients if we change the units of the offset variable. (Note, adding $\log(10)$ on the log scale is multiplying by 10 on the original scale.)

Your Answer	Score	Explanation
<input checked="" type="radio"/> The coefficient estimate is unchanged	✓ 1.00	
<input type="radio"/> The coefficient estimate is multiplied by 10.		
<input type="radio"/> The coefficient estimate is divided by 10.		
<input type="radio"/> The coefficient is subtracted by $\log(10)$.		

Total 1.00 / 1.00

Question Explanation

Note, the coefficients are unchanged, except the intercept, which is shifted by $\log(10)$. Recall that, except the intercept, all of the coefficients are interpreted as log relative rates when holding the other variables or offset constant. Thus, a unit change in the offset would cancel out. This is not true of the intercept, which is interpreted as the log rate (not relative rate) with all of the covariates set to 0.

Question 6

Consider the data

```
x <- -5:5  
y <- c(5.12, 3.93, 2.67, 1.87, 0.52, 0.08, 0.93, 2.05, 2.54, 3.87, 4.97)
```

Using a knot point at 0, fit a linear model that looks like a hockey stick with two lines meeting at $x=0$. Include an intercept term, x and the knot point term. What is the estimated slope of the line after 0?

Your Answer	Score	Explanation
<input checked="" type="radio"/> 1.013		1.00
<input type="radio"/> -0.183		
<input type="radio"/> -1.024		
<input type="radio"/> 2.037		
Total	1.00 / 1.00	

Question Explanation

```
z <- (x > 0) * x  
fit <- lm(y ~ x + z)  
sum(coef(fit)[2:3])  
  
## [1] 1.013
```

