

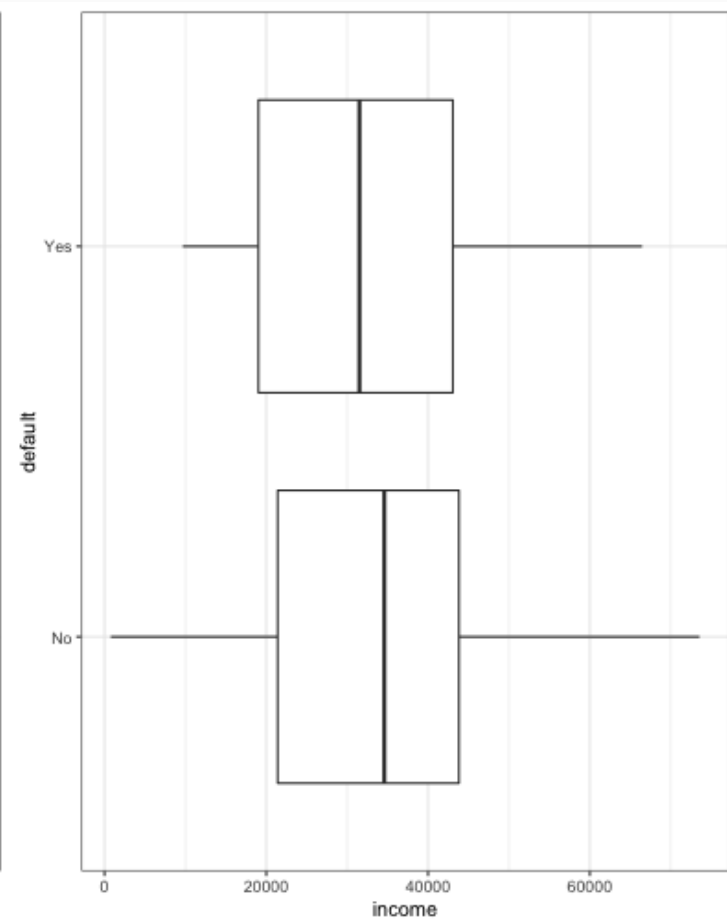
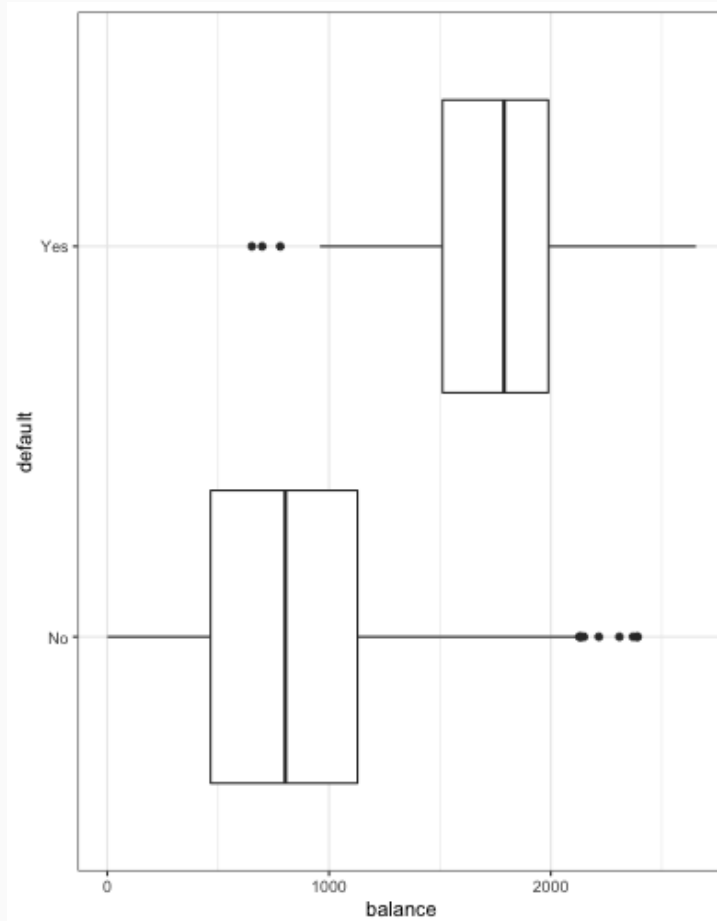
Logistic Regression

Example: Credit Default

```
library(ISLR)
data(Default)
head(Default)
```

##	default	student	balance	income
## 1	No	No	729.5265	44361.625
## 2	No	Yes	817.1804	12106.135
## 3	No	No	1073.5492	31767.139
## 4	No	No	529.2506	35704.494
## 5	No	No	785.6559	38463.496
## 6	No	Yes	919.5885	7491.559

Exploratory Data Analysis



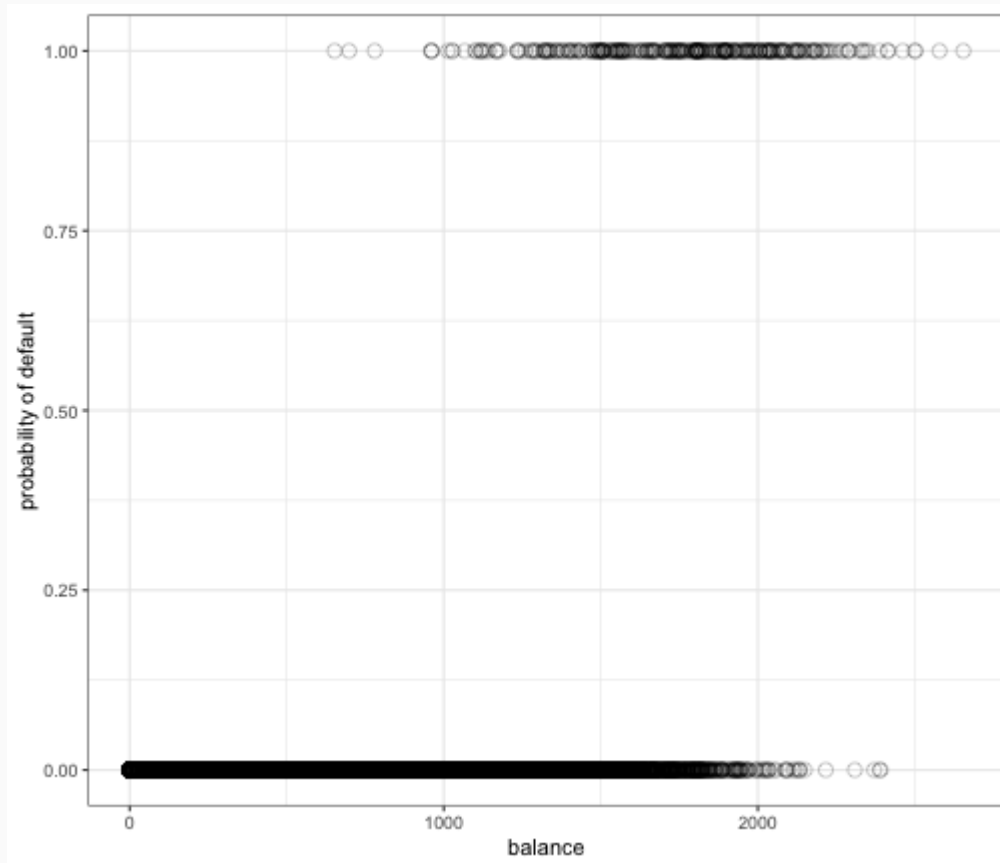
Model Fitting

```
m1 <- glm(default ~ balance,  
           data = Default,  
           family = binomial)  
coef(m1)
```

```
##      (Intercept)          balance  
## -10.651330614      0.005498917
```

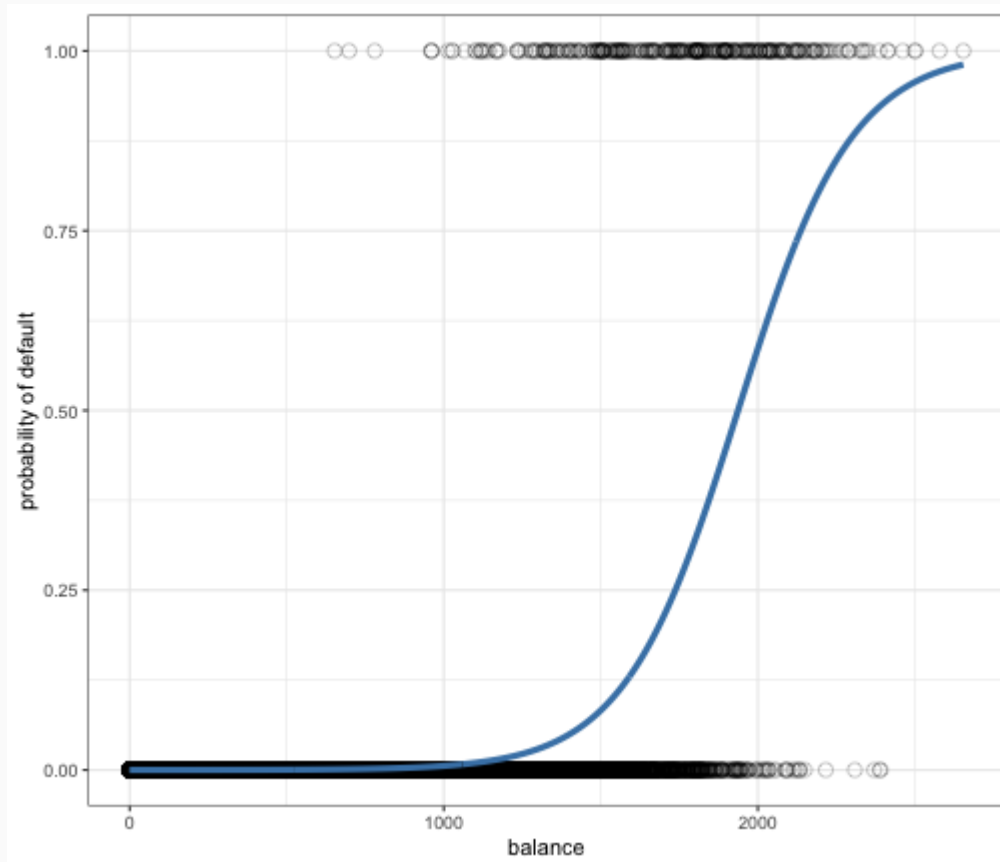
Logistic Model

$$P(Y = 1|X = x_i) = \frac{1}{1 + e^{-(-10.65+0.0055x_i)}}$$



Logistic Model

$$P(Y = 1|X = x_i) = \frac{1}{1 + e^{-(-10.65+0.0055x_i)}}$$



Logistic Model Coefficients

```
summary(m1)$coef
```

##	Estimate	Std. Error	z value	Pr(> z)
## (Intercept)	-10.651330614	0.3611573721	-29.49221	3.623124e-191
## balance	0.005498917	0.0002203702	24.95309	1.976602e-137

Where did those SEs come from?

The Likelihood Function

48 male bank supervisors were asked to assume the role of the personnel director of a bank and were given a personnel file to judge whether the person should be promoted to a branch manager position. The files given to the participants were identical, except that half of them indicated the candidate was male and the other half indicated the candidate was female. These files were randomly assigned to the supervisors. For each supervisor we recorded the gender associated with the assigned file and the promotion decision.

	promoted	not promoted
male	18	6
female	14	10

Is this data consistent with the claim that females are unfairly discriminated against in promotion decisions? What statistical method would you use to make that determination?

A model for promotion

	promoted	not promoted	p(promoted)
male	18	6	$18/24 = .75$
female	14	10	$14/24 = .58$

1. Each decision was independent.
2. All males were promoted with the same probability p_M .
3. All females were promoted with the same probability p_F .

$$Y \sim \text{binomial}(n = 24, p = p_M)$$

$$X \sim \text{binomial}(n = 24, p = p_F)$$

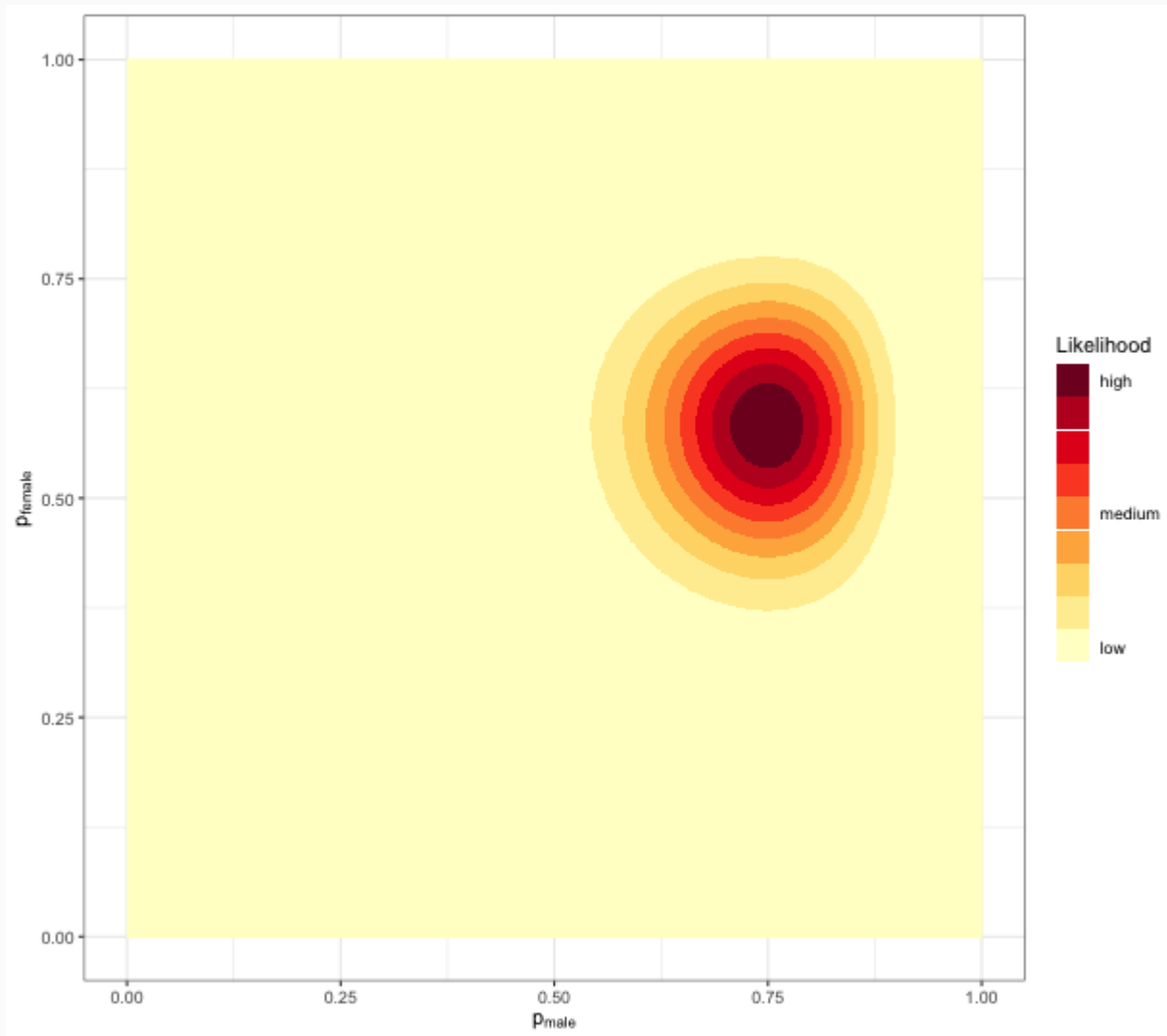
From Probability to Likelihood

$$P(\mathbf{y}, \mathbf{x} | n, p_M, p_F) = \binom{n}{y} p_M^y (1 - p_M)^{n-y} \binom{n}{x} p_F^x (1 - p_F)^{n-x}$$

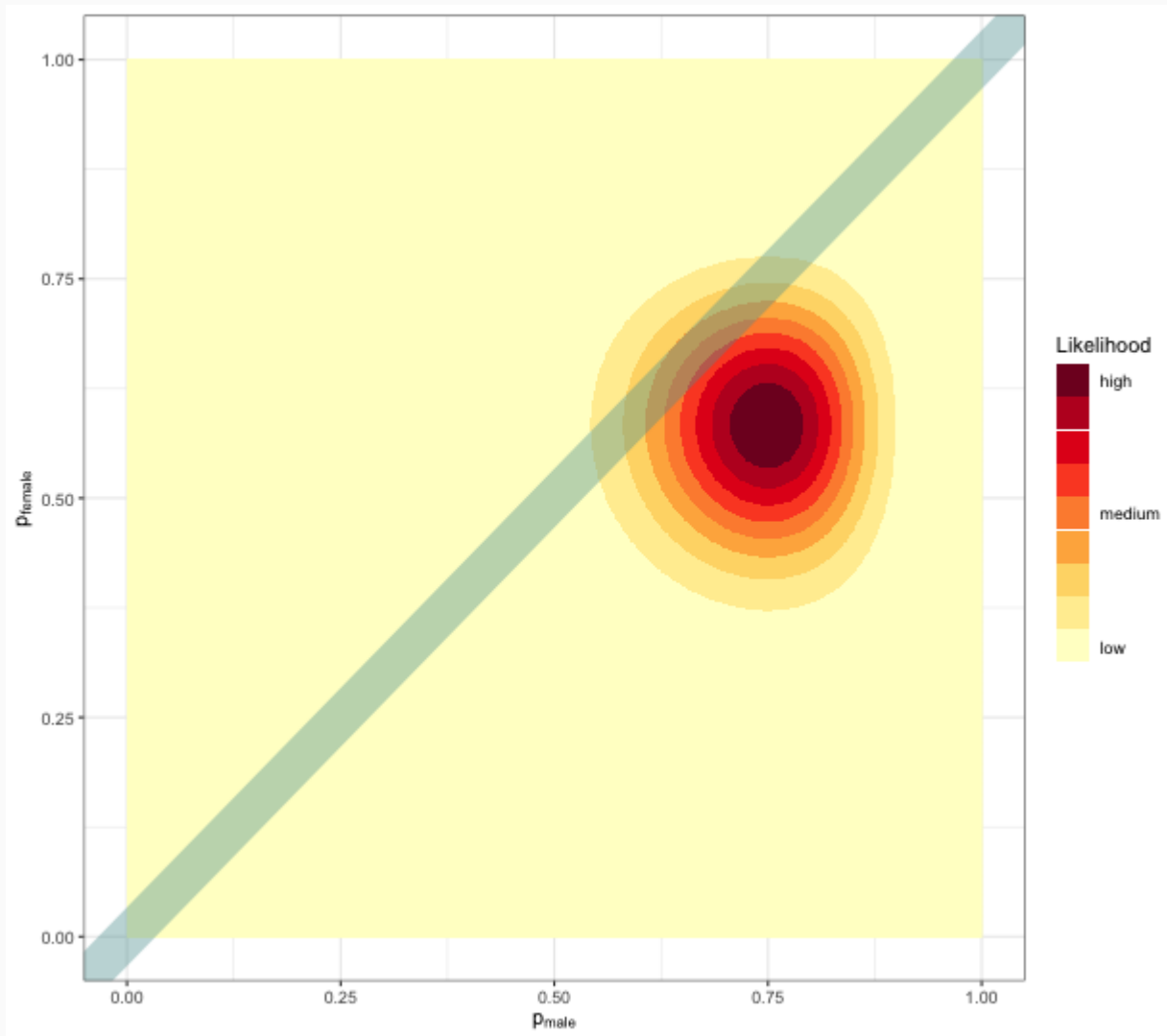
vs.

$$L(\mathbf{p}_M, \mathbf{p}_F | n, y, x) = \binom{n}{y} p_M^y (1 - p_M)^{n-y} \binom{n}{x} p_F^x (1 - p_F)^{n-x}$$

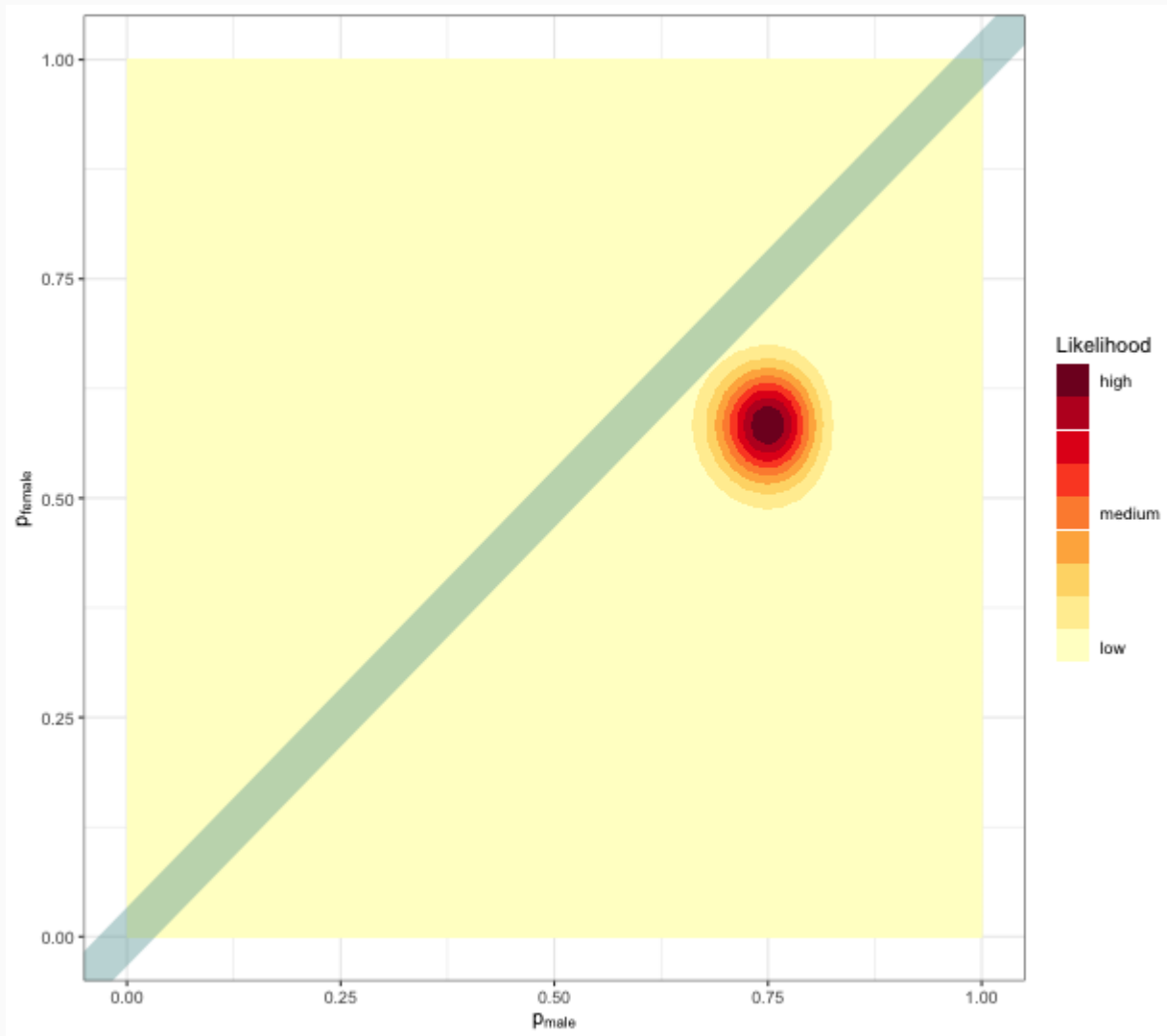
The Likelihood Function



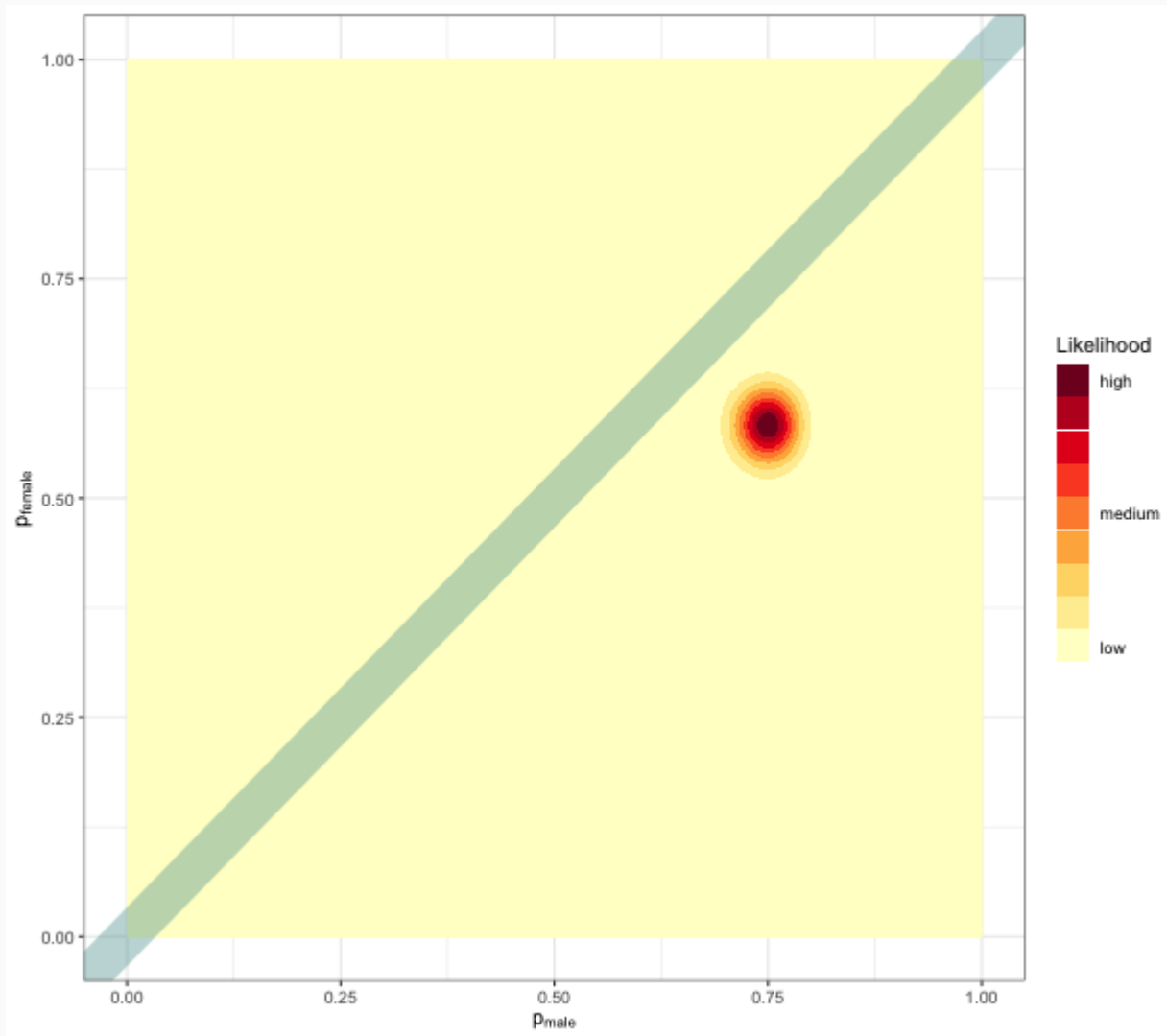
The Likelihood Function



Likelihood Function, more data

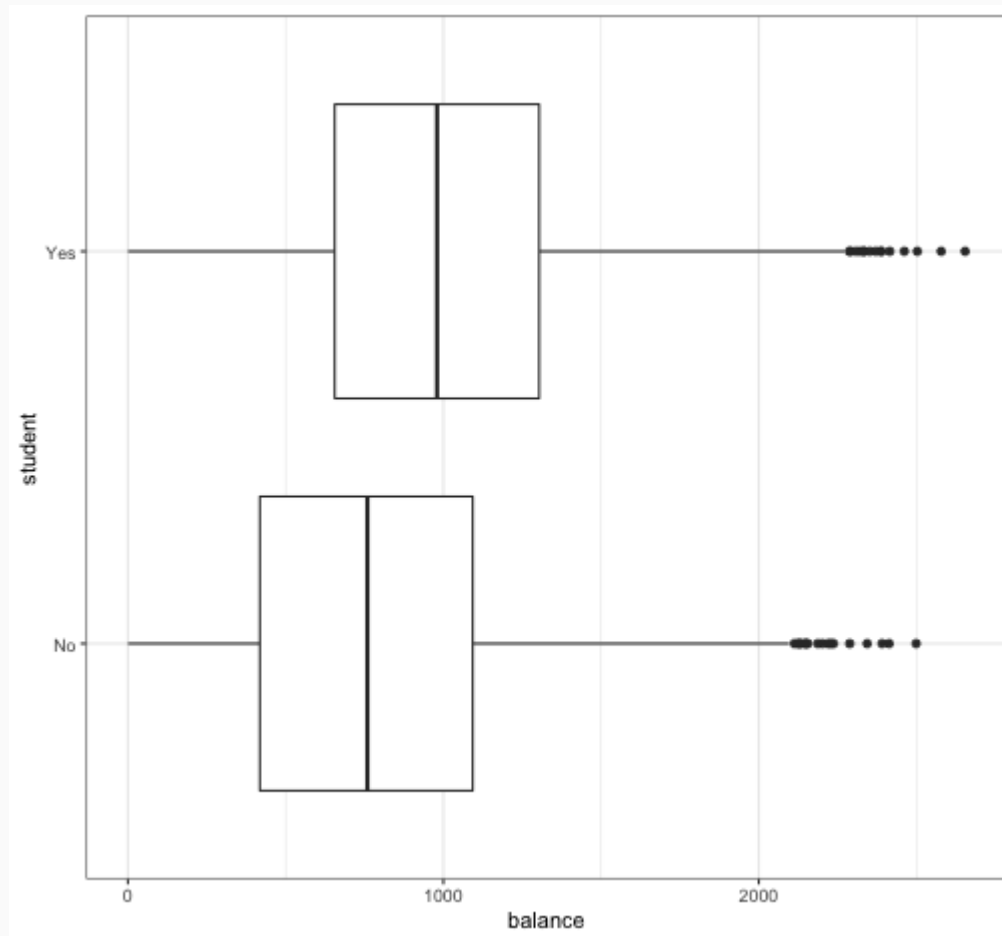


Likelihood Function, even more data



Multiple Logistic Regression

Add student as a predictor?



Multiple Logistic Model

```
m2 <- glm(default ~ balance + student,  
           data = Default,  
           family = binomial)  
summary(m2)$coef
```

##		Estimate	Std. Error	z value	Pr(> z)
##	(Intercept)	-10.749495878	0.369191361	-29.116326	2.230782e-186
##	balance	0.005738104	0.000231847	24.749526	3.136911e-135
##	studentYes	-0.714877620	0.147519010	-4.846003	1.259734e-06

What's going on?

Multiple Logistic Model, cont.

