# Graded Team Assignment: Gender Discrimination Lawsuit

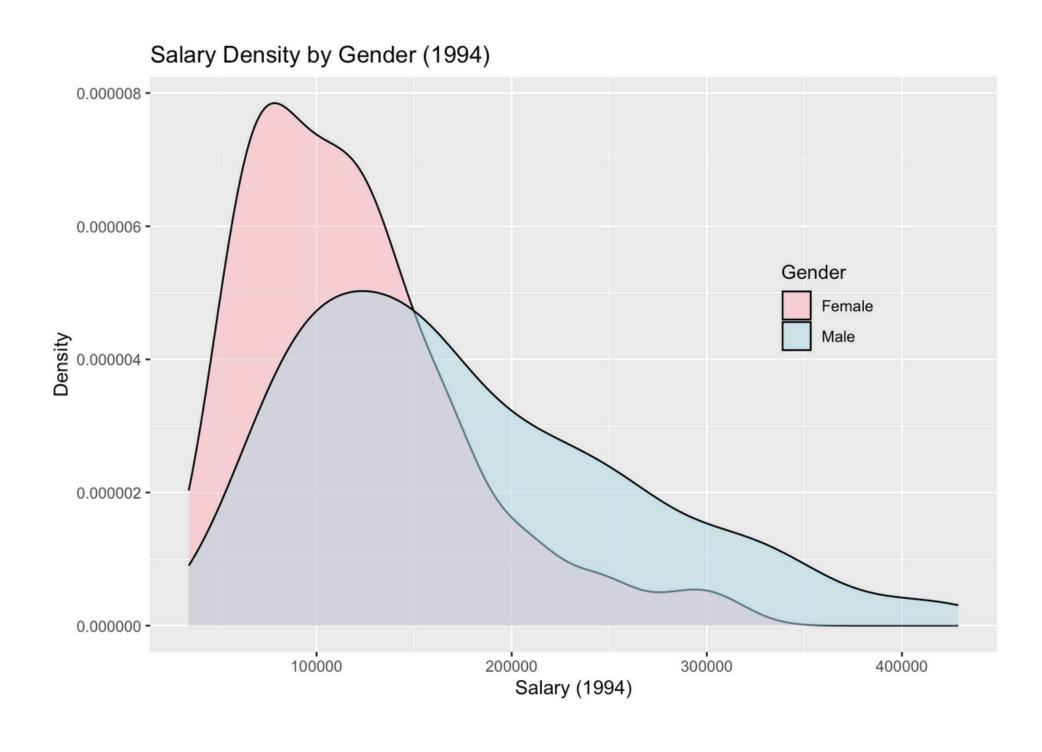
Group 7

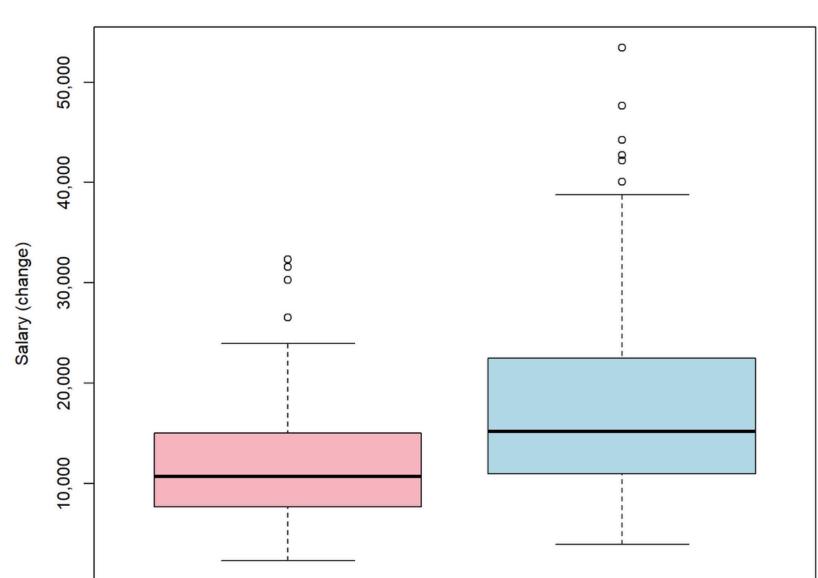
Delwin Ramses van Strien Jo Bum Joon Nguyen Ngoc Thien Huong Zhang Jianyu Zhang Ning Zhao Hengjie

# Key Claims

Claim	Salary Disparity	Promotion Gap		
Argument	Females at the school earn less money than men, on average.	Female faculty at the school are less likely to be full professors, more likely to be assistant professors.		
Target Data	Salary 1994, Salary 1995	Rank		

#### Females Earn Less and Receive Smaller Raises than Men





Salary Change Distribution by Gender

Female salaries are concentrated at lower levels, with a pronounced peak, while male salaries are more spread out and tend to be higher.

Males received higher and more variable raises compared to females, with the median male raise matching the female upper quartile.

Gender

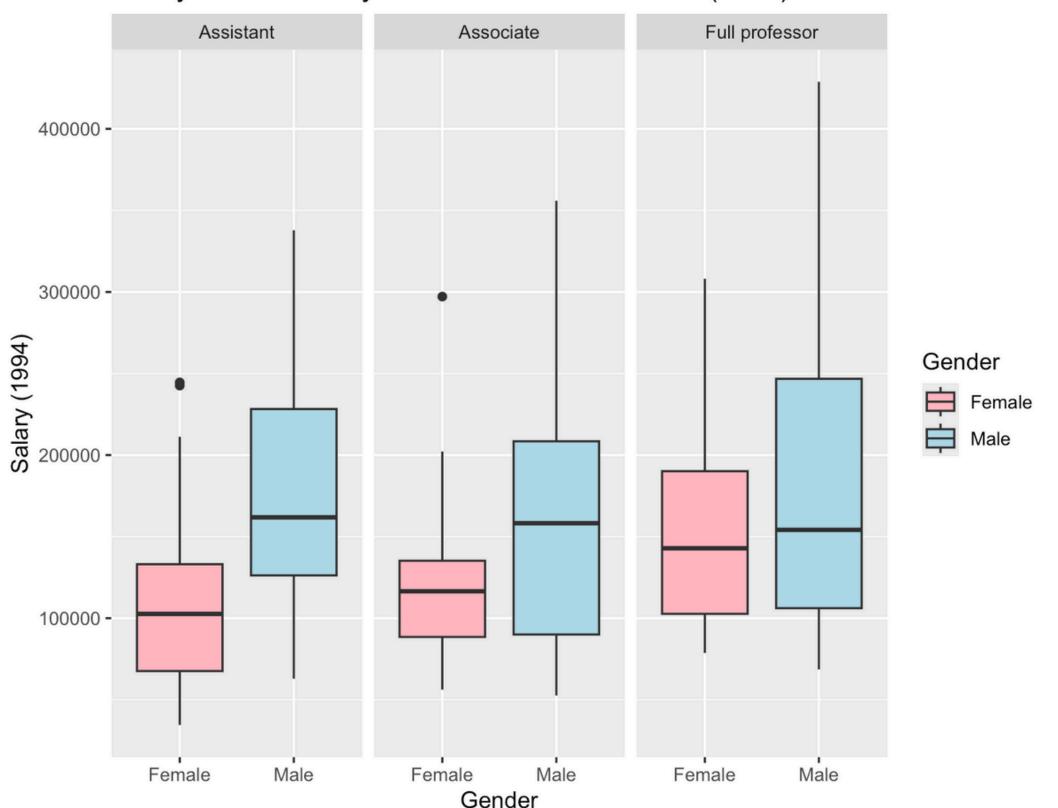
Female

Delwin Ramses van Strien, ZhangNing

Male

### Females Earn Less Than Males at Every Rank

Salary Distribution by Gender within Each Rank (1994)



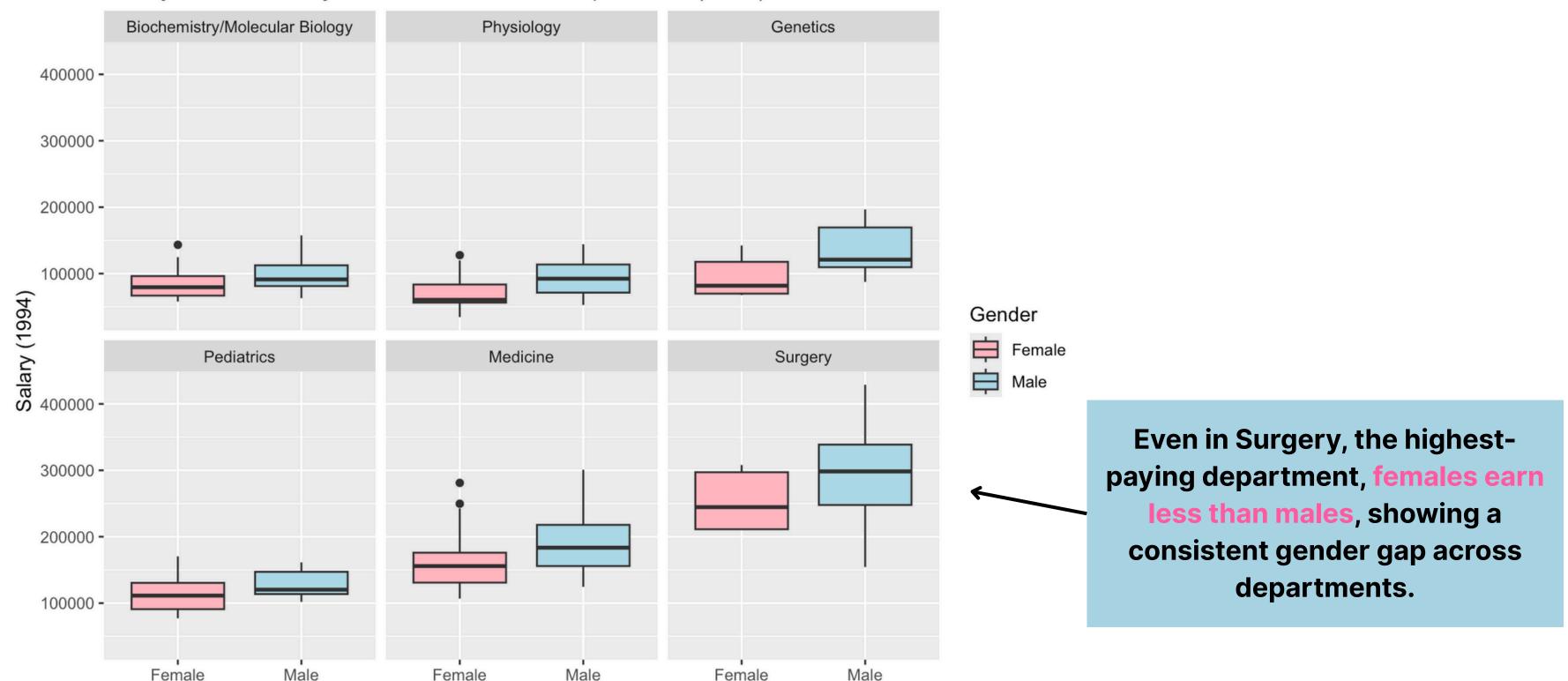
Across all ranks, females earn less than males, even at the top 25%.

The gap is largest among assistant professors, where even the lowest male salaries are near the top of the female range.

# Females Consistently Earn Less Than Males Across All Departments

Salary Distribution by Gender within Each Department (1994)

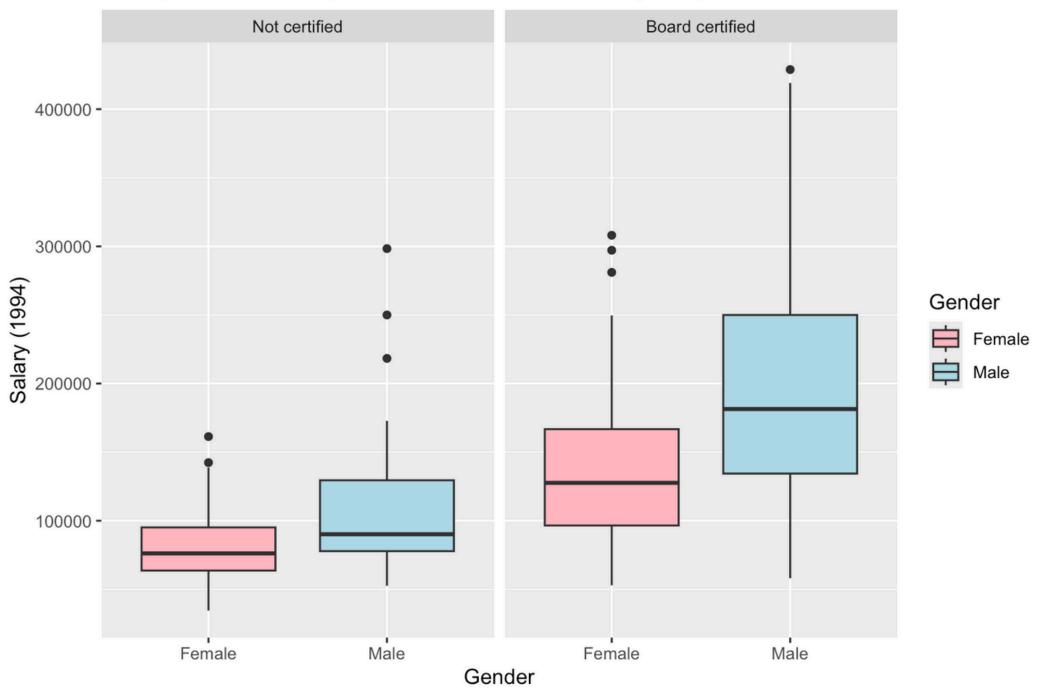
Gender



Nguyen Ngoc Thien Huong, Jo Bum Joon

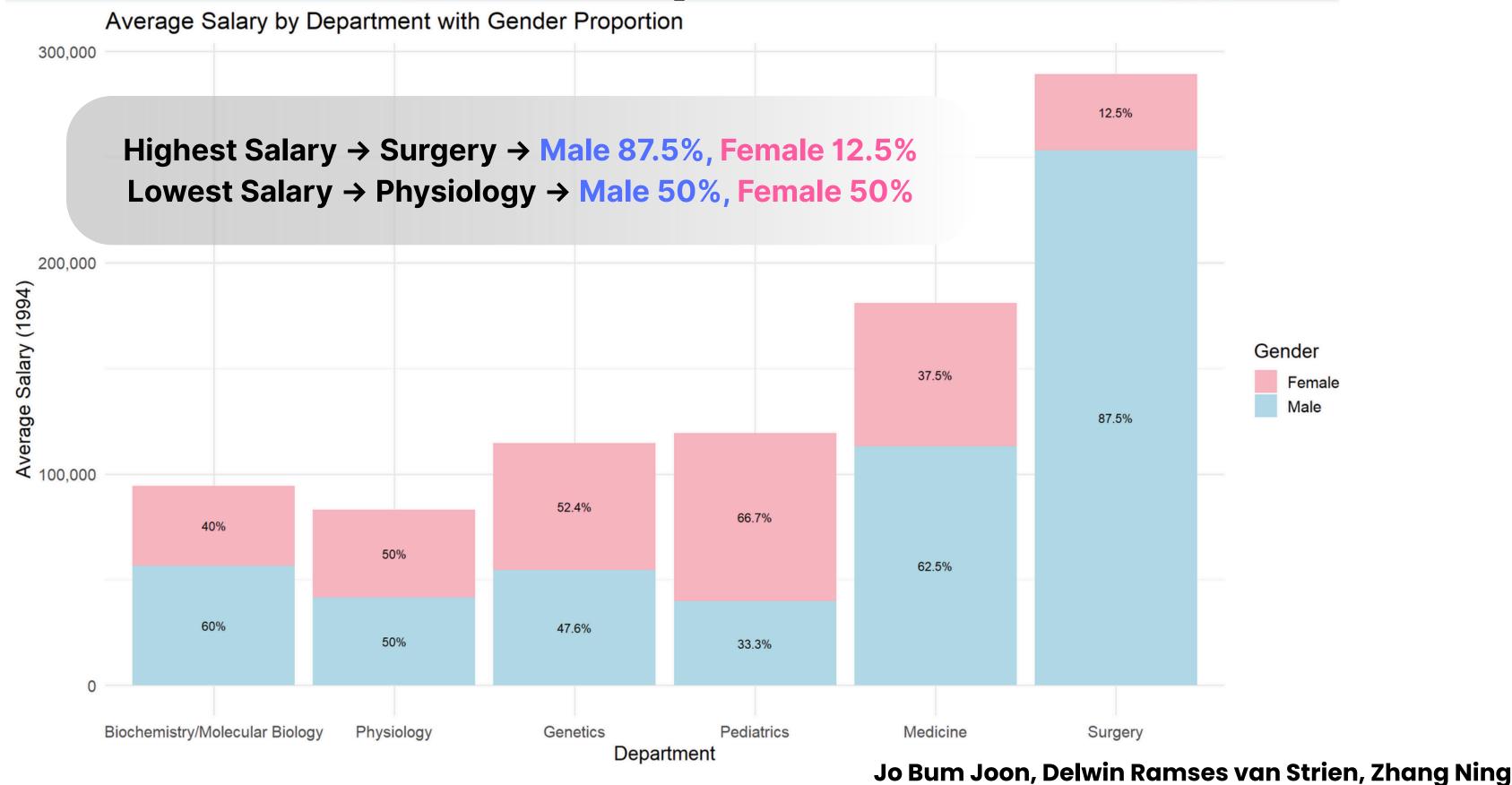
#### Females Are Paid Less Than Males With the Same Certification

Salary Distribution by Gender and Certification (1994)



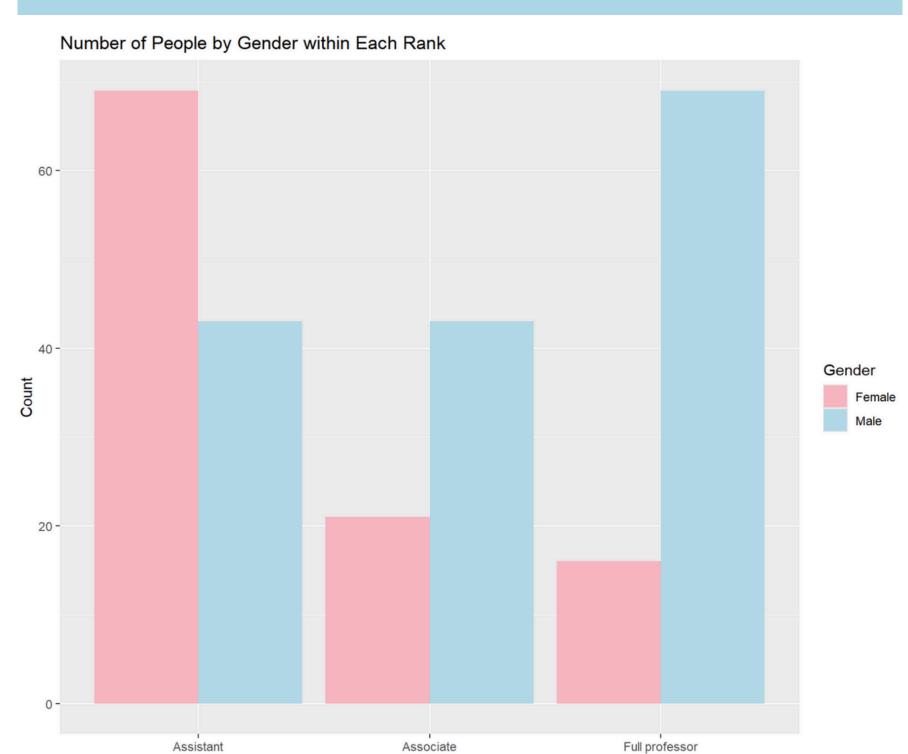
The salary gap widens substantially for board-certified employees, with males earning much more than females.

# Surgery is the Highest-paying Department but has the Lowest Proportion of Females



# Males are More Likely to be Full Professors, resulting in Higher Average Salaries

There are far more female assistant professors than female full professors, while for males the opposite is true.

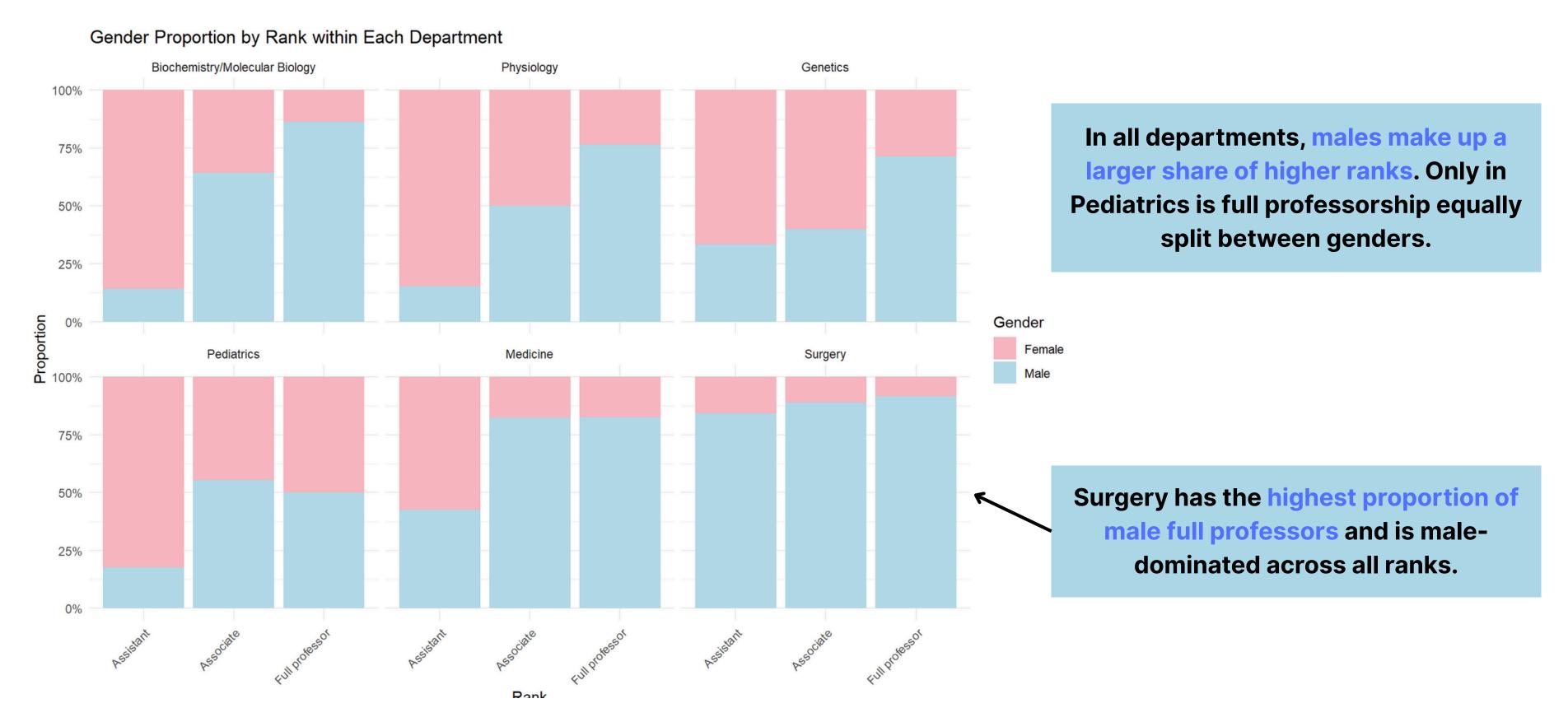


Rank

The proportion of males is greatest at the full professor level, whereas females are more likely to be assistant professors.

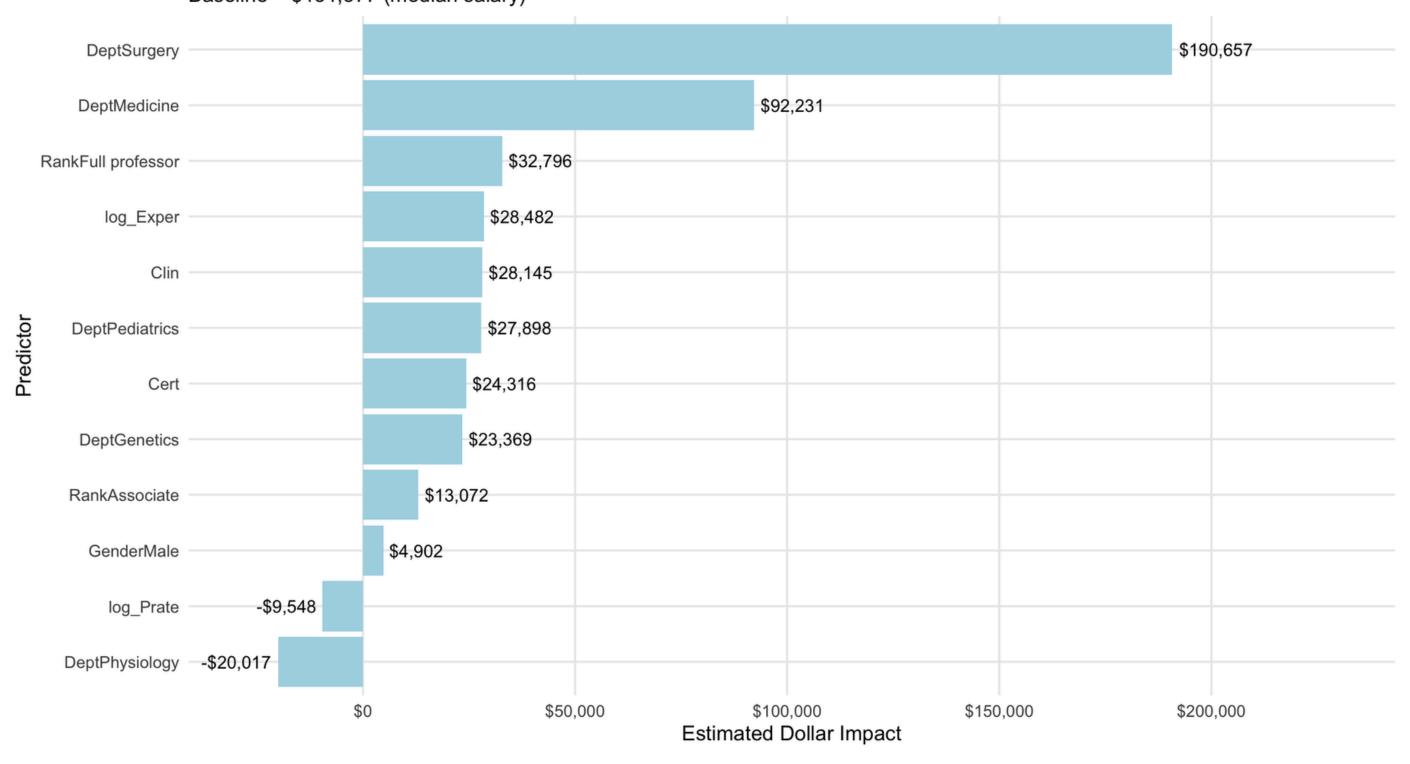


## Gender Gap in Professorship Across Departments



### No Direct Gender Impact on Salary, but Investigation on Department is Needed

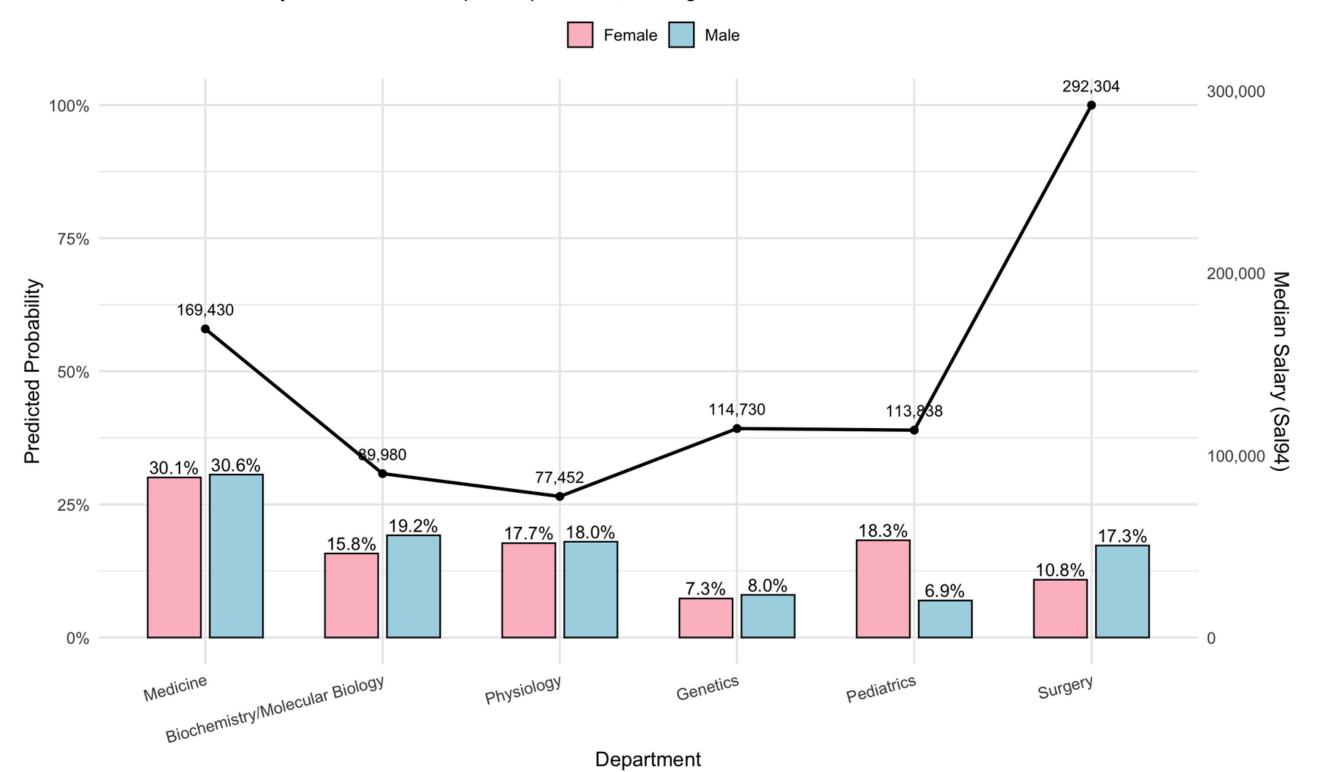
Estimated Dollar Impact on Salary (vs. baseline)
Baseline = \$131,377 (median salary)



# Males Have Higher Chances of Entering Departments with Lucrative Expected Salaries

Males vs Females: Predicted Probability by Department

Males are more likely to enter more well-paid departments, holding other factors constant

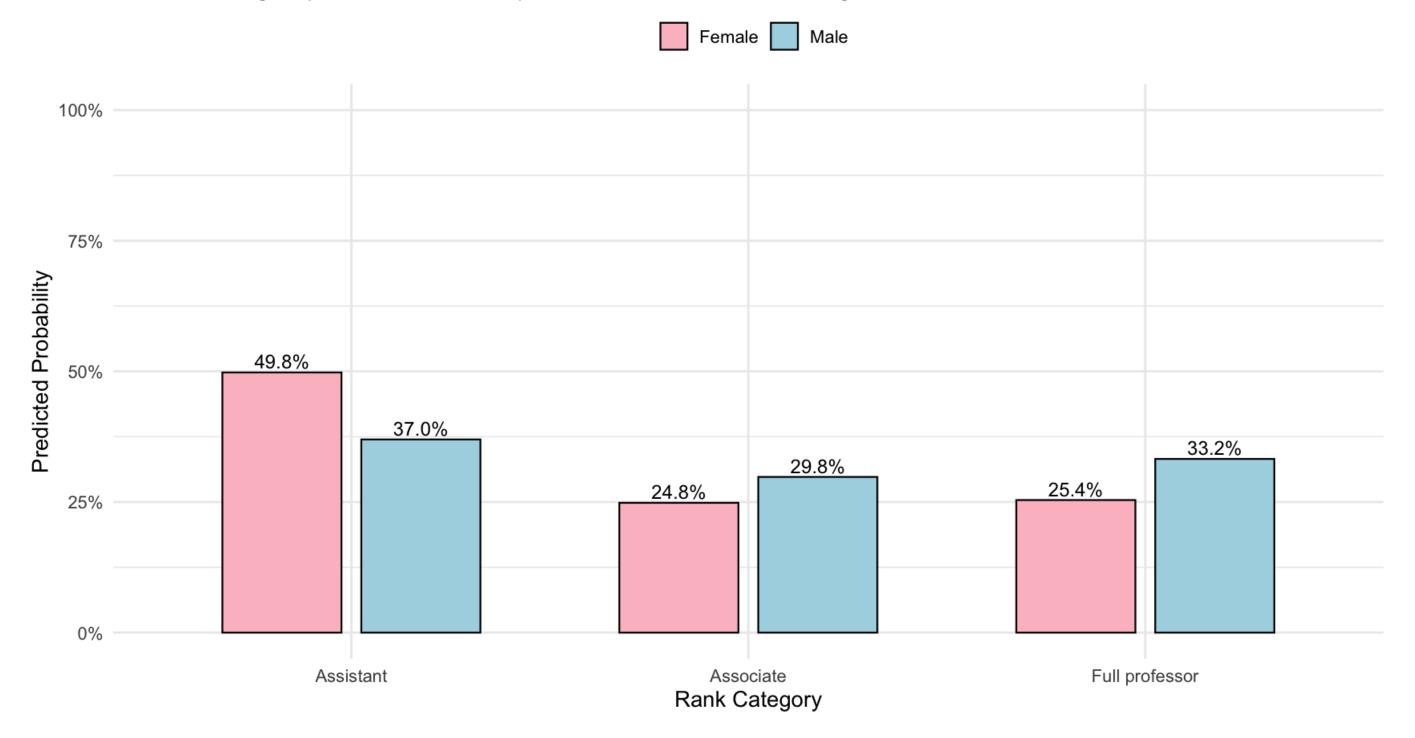


Overall accuracy: 60.9%

# Climbing the Ladder: Males Have Higher Chances of Reaching Higher Academic Ranks

Males vs Females: Predicted Probability of Each Academic Rank

Males retain higher predicted chances of promotion to senior ranks, holding other factors constant



Overall accuracy: 72.8%

## Conclusions

#### **Salary Analysis**

- Descriptive comparisons show clear pay differences between genders.
- Regression analysis, which controls for potential confounding factors, further confirms that female doctors' salaries remain consistently lower than male doctors' salaries.



#### **Promotion Perspective**

- Statistical modelling of career progression and visualization of hierarchical outcomes reveal that women are substantially underrepresented in higher ranks.
- The likelihood of female doctors advancing up the corporate ladder is evidently lower than for their male counterparts.

The consistency across multiple methods strengthens the reliability of this conclusion

**GENDER DISCRIMINATION** is present in the environment of female doctors

# Appendix

# Salary Levels (Linear Regression Model)

Model	Constant	Gender	Clin	Cert	log_Prate	log_Exper	Physiology	Genetics	Pediatrics	Medicine	Surgery	Associate	Full Professor	R <sup>2</sup>	Adj. R²
m1: Sal95 ~ Gender + Dept + Rank + Exper + Prate + Cert + Clin	15621.2	-297.4	16695.8	17152.1***			-10347	22746.1*	21712.2	74331.6***	174875***	16335.2**	34366.8***	0.9027	0.8967
m2: Log_Sal94 ~ Gender + Clin + Cert + log_Prate + log_Exper + Dept + Rank	10.46***	0.04	0.19***	0.17***	-0.08	0.20***	-0.17***	0.16***	0.19**	0.53***	0.87***	0.09**	0.22***	0.931	0.927

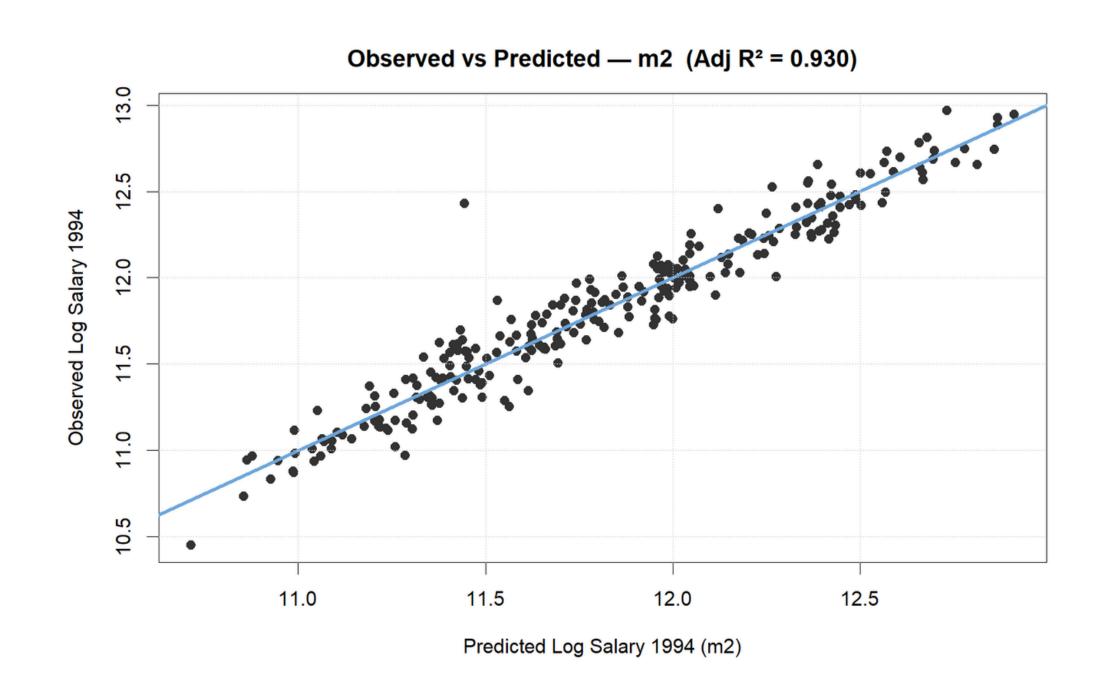
<sup>\*</sup> represents the level of significance: \* p<.05, \*\* p<.01, \*\*\* p<.001; log-model %  $\approx$  100·(exp( $\beta$ )-1)

# What the Linear Regression Models Show (m0-m2)

- m1 (+ Dept + Rank + Exper + Prate +
   Cert + Clin): Gender n.s.; Experience ≈
   +\$3.3k/year; large Department and
   Rank effects (R² ≈ 0.90).
- m2 (perform log transformation):
   Gender n.s.; strong Dept/Rank and log-Experience effects (R<sup>2</sup> ≈ 0.93).

#### **Key Takeaways:**

- The overall gap is mostly explained by department & rank composition.
- The model achieved high reliability.

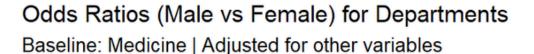


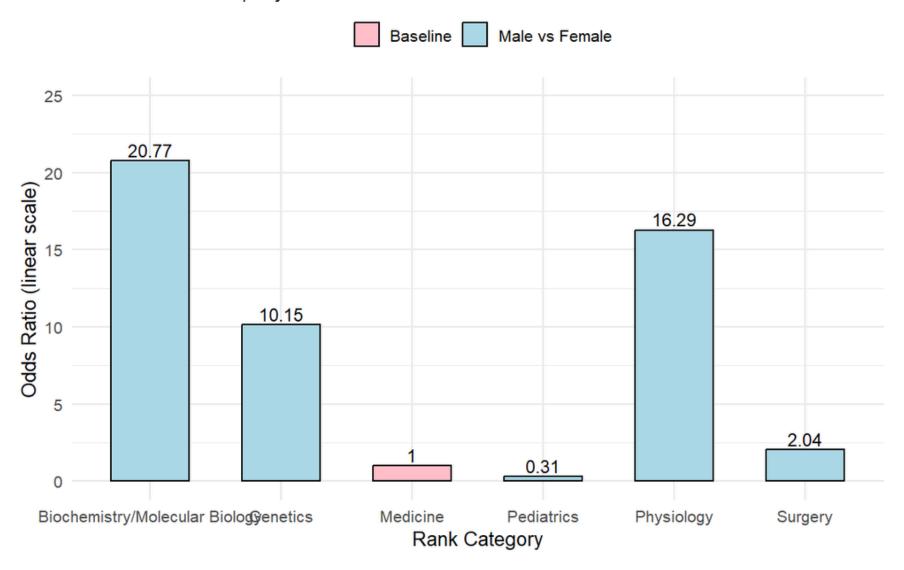
# Logistic Regression Result - Department

Coefficients: (Intercept) GenderMale RankAssociate RankFull professor Biochemistry/Molecular Biology -67.203022 3.0332770 0.63693807 3.2275770 Physiology -68.480742 2.7903633 0.82501673 3.2015326 Genetics -48.934490 2.3177236 -0.51543370 1.8224048 Pediatrics 8.277456 -1.1649664 0.49637354 -0.872353917.517177 0.7126758 -0.03703428 -0.7480315Surgery ClinPrimarily clinical emphasis CertBoard certified Prate Exper Biochemistry/Molecular Biology 19.237367 -5.2701351 10.653185 -0.01193282 Physiology 21.316542 -4.9550763 10.851131 -0.07056389 Genetics 15.345054 -4.0331756 8.064849 -0.04321957 0.3349324 -1.448437 -0.02417956 Pediatrics -4.123047 -7.264527 1.5846128 -4.339547 0.07049989 Surgery

Multi-logistic regression output

## Logistic Regression Result - Department





#### Odds ratio comparison against baseline group (Medicine)

Using **Department** as the target variable and the rest as predictor variable, we found that: Keeping other variables constant, being a male increase the relevant chance of joining **all other departments** except pediatrics department.

## Logistic Regression Prediction Logic

Original Table

Male Table

Female Table

Gender	Cert	Exper
Male	1	8
Female	0	9
Male	0	7
Female	1	2

Gender	Cert	Exper
Male	1	8
Male	0	9
Male	0	7
Male	1	2

Gender	Cert	Exper
Female	1	8
Female	0	9
Female	0	7
Female	1	2

Given this hypothetical dataset, we manipulate the dataset to create **2 new ones** to assume that one are all males and one are all females, so that we keep all other variables constant and use our logistic regression model to predict the chance that male or female can be promoted to higher positions.

## Logistic Regression Result - Rank

```
Coefficients:

(Intercept) GenderMale DeptPhysiology DeptGenetics DeptPediatrics DeptMedicine DeptSurgery
Associate -1.118726 0.9443212 0.9285269 0.02587890 0.2214147 -0.3956320 -1.533060
Full professor -2.668739 1.1681863 0.9621182 0.05475274 -1.2262315 -0.7718433 -2.590581
ClinPrimarily clinical emphasis CertBoard certified Prate Exper
Associate -0.8738808 -1.2807891 -0.4047080 0.4516098
Full professor -0.7515261 -0.5936204 -0.3799801 0.5414052
```

Multi-logistic regression output

### Logistic Regression Result - Rank

Odds Ratios (Male vs Female) for Academic Rank Baseline: Assistant | Adjusted for Dept, Clin, Cert, Prate, Exper Baseline Male vs Female Odds Ratio (linear scale) Assistant Associate Full professor Rank Category

Odds ratio comparison against baseline group (Assistant)

Using **Rank** as the target variable and the rest as predictor variable, we found that: Keeping other variables constant, being a male increase the relevant chance of becoming an associate professor by about **1.5 times**, and are more than **2 times** likely to become a full professor.