Oregon Institute of Technology

# Faculty Pay Equity Analysis

#### **Report of Findings**

February 21, 2023

Privileged and Confidential—Do Not Copy or Distribute. Document only to be reviewed in conjunction with Segal commentary. © 2023 by The Segal Group, Inc.



Segal Introductions

**Moshe Mayefsky** VP, Senior Consultant **Greg Saylor** Associate Consultant

#### Keegan O'Boyle Associate

National consulting firm with 1,100+ employees, celebrating over 80 years of serving clients

Mission-driven: **Providing trusted** advice that improves lives

Independent, objective, and **employee-owned** 

Not any solution — **your solution**; personalized advice and help



# Report of Findings

#### **1. Summary of Findings**

- 2. Methodology
- 3. Detailed Analysis



#### Background

• Oregon Institute of Technology engaged Segal to conduct a pay equity study to understand the impact gender and race may have on compensation for its faculty.

#### **Objectives**

- Determine the extent to which there may be a systemic bias in pay with respect to gender and/or race/ethnicity, while controlling for effects of other variables, such as education, experience, and responsibilities of the job.
- To accomplish these objectives, Segal conducted a comprehensive and independent statistical analysis of multiple variables and their relationship to pay. The focus was to determine the primary drivers of pay and to help understand the impact gender and/or race may have on pay levels.



### **Systemic Findings**

- The following variables proved to be primary drivers of faculty compensation:
  - Rank Campus Department
  - Division Time in Rank
- The primary drivers of pay above represent reasonable characteristics that would be expected to contribute to pay in the market.
- The data analyses and predictive model conclude that there **do not appear to be systemic inequities based on gender and/or race/ethnicity** after the above predictive variables were considered.



#### **Individual Findings**

• There were 10 individual outliers (6.3% of the population), as can be expected, with total salaries below or above 2.0 standard errors from the predictive model. These outliers should be examined on a case-by-case basis, and other reasonable factors influencing their pay (such as prior experience not considered, performance, etc.) should be noted for each individual.

	Count	Percentage of Cohort Group
Population	160*	100.0%
Low Outliers	3	1.9%
Female / Male	0/3	0.0% / 1.9%
Minority / Non-Minority / Not Disclosed	0/3/0	0.0% / 1.9% / 0.0%
High Outliers	7	4.5%
Female / Male	4/3	2.5% / 1.9%
Minority / Non-Minority / Not Disclosed	1/6/0	0.6% / 3.8% / 0.0%

#### **Outliers by Gender & Race/Ethnicity**

\* Includes bargaining unit members and department chairs (only the base salary)



**Next Steps and Recommendations:** 

1 Complete review of low individual outliers

**2** Implementation of individual pay adjustments, as appropriate

Bolster data-upkeep



Incorporate results into current pay policies and practices

- Continue reviewing individual outliers to ensure rationale behind pay differences are captured, where appropriate.
- Determine the extent of pay adjustments for low outliers and potential freezes for high outliers. Decide on a timeframe for any pay adjustments.
- Continue to collect relevant data to all components of pay (specifically surrounding grant funded positions) to ensure completeness of data and to improve future pay equity studies.
- Utilize the predictive models to assess compensation levels when hiring or promoting to maintain equity.



# Report of Findings

- **1. Summary of Findings**
- 2. Methodology
- 3. Detailed Analysis

### Pay Equity Analysis Framework







- Collect quantitative data including categorical, discrete, and continuous variables.
- Determine appropriate number and types of employee groupings.
- Identify initial pay gaps within each group, considering a variety of variables.
- Determine data transformations necessary for more rigorous modeling.

To ensure the data conformed to a normal (i.e., bell-shaped) distribution, we used the natural log of the salaries as the dependent variable in the multiple regression analysis.





- Identify variables that have the largest influence on pay differentiation, while accounting for multiple factors.
- Determine significant differences across groups while controlling for effects of other significant variables.
- Develop a predictive model based on the primary drivers of pay.
- Conventional standards used to measure explanatory power (R<sup>2</sup>) and variable significance (p-value).



The Adjusted R<sup>2</sup> value for our predictive model is **87%** suggesting that the set of predictor variables does a strong job of estimating variances in salaries.





- Determine the extent to which there is systemic pay inequity potentially stemming from a gender or race/ethnicity bias.
- Compare actual pay to expected pay for each employee and provide a list of individual outliers.
- Calculate associated costs necessary to remediate any issues under various remediation strategy alternatives.

No systemic inequities based on gender and/or race/ethnicity were found. Individual outliers should be reviewed on a case-by-case basis to determine potential pay adjustments.



#### **Limitations of Pay Equity Studies**



Due to these limitations, and since pay can differ for each employee for these and many other individualized reasons, it is impossible for any pay equity study to account for all differences in pay.



# Report of Findings

- 1. Summary of Findings
- 2. Methodology
- 3. Detailed Analysis



### Overview of Data

#### **Overview of Data Elements Used in the Study**

• The following data fields were provided and considered for inclusion in the statistical model:

Data Fields			
Unique Employee ID	Hire Date	Supervisor	
Name	<ul> <li>FTE Percentage</li> </ul>	• DOB	
ECLS Code	<ul> <li>Annual Appointment</li> </ul>	Ethnicity Code Description	
Location	<ul> <li>Annual Salary</li> </ul>	Ethnicity IPEDs Description	
Division	Current Tenure	Gender	
Department	Current Rank	<ul> <li>Degree (partial)</li> </ul>	
Job Title	Rank History	COLA History	
Position Number	Current Hire Date	Department Chair Status	



### Descriptive Statistics by Minority Status

#### Average Salary by Minority Status



\* Totals exclude one (1) individual who declined to respond to the Race/Ethnicity question

### Descriptive Statistics by Gender

Average Salary by Gender



#### **Composition by Rank**



16

🔆 Segal

### Predictive Model

#### **Multivariate Regression Analysis Results**

Predictive Power of Each Variable (Adjusted R<sup>2</sup>)



- The predictive model provides an overall Adjusted R<sup>2</sup> value of **87%.** This suggests that the set of predictor variables does a strong job estimating the variance in salaries overall.
- Rank and Division are the most significant factors influencing pay, followed by Department, Campus and Time in Rank.
- "Other Factors" that can influence pay may include:
  - Rank at hire
  - Performance
  - Other data not captured in the HRIS system

Gender and races/ethnicities were not noted as a statistically significant variables after the above predictive variables were considered.



### Predicted Salary vs. Standard Errors From Predicted Salary



🔆 Segal 18

### Predicted Salary vs. Standard Errors From Predicted Salary



Segal 19

### Conclusion

#### Based on the analyses above, we conclude:

- Oregon Institute of Technology uses a reasonable set of data fields to differentiate pay, as determined through the development of a strong predictive model.
- There do not appear to be systemic inequities based on gender and/or race/ethnicity after other predictive variables were considered.
- As can be expected, 10 individual outliers were identified. These outliers should be examined on a case-by-case basis, for other reasonable factors influencing their pay.

