## Measuring Pay Equity: Small Sample Size Individual Analysis

Individual-level analyses in pay equity investigations have two goals:

- 1) identify individuals who may be negatively impacted by potential pay disparities,
- 2) estimate the amount by which they may be negatively impacted.

When sample size is small and the 30/5-rule criteria cannot be met for regression analysis, individual-level analysis is still possible with data modeling approaches, but the results must be interpreted with caution and additional care.

There are many different methods to individual-level analysis. Here is a general analytical framework that can be applied to measure pay equity between individuals, including small sample size situations.

Overall, this approach involves two general steps:

- 1) Step 1--Estimate Predicted Pay
- 2) Step 2--Compute Individual Pay Difference

Please note that the method and steps detailed in this tool is not authoritative or definitive. The purpose of this tool is to provide a generic analytical framework that employers may use as a starting point and adapt and refined to the unique needs of their organization.

When sample size is small, it is not possible to complete a regression analysis among individuals performing substantially similar work to compute predicted pay. However, it is possible to compute an estimate of individual pay difference in standard deviation (SD) units, which is critical to identifying potentially underpaid individuals. The following provides an outline the steps to compute standardized pay difference:

## Step 1: Estimate Quasi-Predicted Pay for Individual Employees

When sample size is small, *quasi*-predicted pay can be obtained from a regression analysis of company wide data. To estimate *quasi*-predicted pay for each employee, specify a regression model of worker and job characteristics but excludes any group indicator variables, (e.g., gender).

For example:

1) Regression Model for Pay Equity:



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It is important to note that *quasi*-predicted pay is not estimated from a group of individuals performing substantially similar work. Therefore, the *absolute* value of quasi-predicted pay does not directly indicate pay (in)equity. The *relative* value of quasi-predicted pay, however, is meaningful and will be used to compute individual standardized pay difference.

### Step 2: Compute the Estimated Pay Differences for each Individual

**Compute Raw Difference:** After quasi-predicted pay is obtained for each individual employee, the next step is to compute the raw difference between *actual*-pay and *quasi-predicted* pay. These raw difference values are not relevant to pay equity analysis on their own, they must first be converted into standardized deviation units, as explained below.

*Standardize Pay Differences by Substantially Similar Work Group*: Once raw differences are computed, the next step is to identify and standardize the raw differences, separately, for each group who are performing substantially similar work. These standardize pay differences are interpretable and can be used to identify negatively impacted individuals.

Once raw differences are standardized by substantially similar work groups, it is possible to identify underpaid individuals even when the groups are small. Negative standardized pay differences indicate individuals who are underpaid.

#### Example:

A very simplified hypothetical example illustrates this process:

**Step 1:** Complete a Regression Analysis on Company Wide Data (excluding group indicator, e.g. gender).

Regression model based on Company Wide Data:

**Quasi-Predicted Pay =** 50,000<sub>Base Salary</sub> + 2,000×**Tenure** + 10,000×**Performance** 

**Step 2**: Compute Quasi-Predicted Pay Using Company Wide Data For the following hypothetical sample of employees with the given levels of Tenure and Performance, guasi-predicted pay is presented in the last column:

Emp ID	SSWG	Tenure	Performance	Quasi- Predicted Pay
49	Maintenance	1	1	\$62,000
30	Maintenance	5	2	\$80,000
79	Maintenance	6	3	\$92,000
46	Production	6	4	\$102,000
48	Production	4	5	\$108,000
62	Production	2	3	\$84,000
33	Production	4	4	\$98,000
51	Production	7	1	\$74,000

**SSWG**=Substantially Similar Work Group

Please note, the absolute value of the quasi-predicted pay values should not be interpreted.

# Step 3: Compute the Estimated Pay Differences for each Individual

The estimated *raw* pay difference for each employee is calculated as the difference between the employee's *actual pay* and their *predicted pay* estimated in Step 1. For the hypothetical sample of employees:

Emp ID	SSWG	Tenure	Performance	Quasi- Predicted Pay	Actual Pay	Raw Difference
49	Maintenance	1	1	\$62,000	\$70,000	\$8,000
30	Maintenance	5	2	\$80,000	\$90,000	\$10,000
79	Maintenance	6	3	\$92,000	\$92,000	\$0
46	Production	6	4	\$102,000	\$65,000	-\$37,000
48	Production	4	5	\$108,000	\$66,000	-\$42,000
62	Production	2	3	\$84,000	\$50,000	-\$34,000
33	Production	4	4	\$98,000	\$45,000	-\$53,000
51	Production	7	1	\$74,000	\$50,000	-\$24,000

**SSWG**=Substantially Similar Work Group

Please note, the absolute value of the raw difference values should not be interpreted.

Step 4: Separately Standardized	Raw I	Difference	for Ea	ach Substantiall	y Similar Working
Groups.					

Emp	SSWG	Tonuro	Porformanco	Quasi-	Actual	Raw	SD
ID	33110	Tenure	Fellomance	Predicted Pay	Pay	Difference	Difference
49	Maintenance	1	1	\$62,000	\$70,000	\$8,000	0.38
30	Maintenance	5	2	\$80,000	\$90,000	\$10,000	0.76
79	Maintenance	6	3	\$92,000	\$92,000	\$0	-1.13
46	Production	6	4	\$102,000	\$65,000	-\$37,000	0.09
48	Production	4	5	\$108,000	\$66,000	-\$42,000	-0.38
62	Production	2	3	\$84,000	\$50,000	-\$34,000	0.38
33	Production	4	4	\$98,000	\$45,000	-\$53,000	-1.41
51	Production	7	1	\$74,000	\$50,000	-\$24,000	1.31

Separately standardize **Raw Differences** for each Substantially Similar Working Group (SSWG). In this example, standardized raw differences are computed for Maintenance employees and separately standardized raw differences are computed for Production employees. It is critical to compute standardized raw differences because the means and standard deviations for each SSWG are unique and different.

*Interpreting SSWG=Maintenance*: In this example, there are three individuals performing Maintenance work. It is important to only interpret the SD Difference results. Employees ID=49 and ID=30 have positive SD Differences—they are overpaid. Employee ID=79, however, is negative (SD=-1.13)—employee ID=79 is underpaid by 1.13 standard deviations.

*Interpreting SSWG=Production.* In this example, there are five individuals performing Production work. It is important to only interpret the SD Difference results. Employees ID=46, ID=51, and ID=62 have positive SD Differences—they are overpaid. The most overpaid is Employee ID=51—he is overpaid by 1.31 SDs. Employees ID=48 and ID=33 have negative SD

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differences. Employee ID=48 is -0.38 standard deviations below, and employee ID=33 is -1.41 standard deviations below.

The Production workers example demonstrates why interpreting Raw Differences can be misleading. In this case, all five production workers have negative raw differences, which may suggest that everyone is underpaid. These *absolute* differences are not meaningful because they are computed from quasi-predicted pay. By standardizing the raw differences, into SD Differences, it is possible to estimate the *RELATIVE* difference in pay between these five production employees. In this case, employee ID=46, ID=62, ID=51 are overpaid and employee ID=38 and ID=33 are underpaid.

#### Next Step

Once the individual pay gaps are computed and the underpaid individuals are identified, the next step is to investigate the cause of that gap. This is often referred to as a "cohort analysis." See XXX (insert reference for cohort analysis) for details on cohort analysis methods