

Pay Ratio: “Other Reasonable Methods”

A Simulation Approach to Estimate Median Employee Pay

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Overview of Proposed Pay Ratio Rules

- The SEC proposed adding Item 402(u) to the executive compensation disclosure rules to implement Section 953(b) of Dodd-Frank
 - Note that this is a different section than the CD&A (i.e., ratio might be disclosed elsewhere)
- "Total Compensation" for the CEO and the median employee calculated using Summary Compensation Table rules for NEOs
- All employees need to be considered in calculating the median including full-time, part-time, seasonal or temporary workers as well as non-US employees
- The SEC is proposing that companies comply with the rule for the first fiscal year after the effective date of the rule
 - For calendar-year companies, likely to be 2016 proxy disclosure for 2015
- To identify the "median employee," proposed Instruction 2 to Item 402(u) would permit the use of:
 - Reasonable estimates to identify median or to calculate elements of total compensation for median
 - Statistical sampling or "other reasonable methods"
 - Other consistently applied compensation measures to identify the "median employee"
 - Once "median employee" is identified, total compensation must be calculated in accordance with Summary Compensation Table rules

Statistical Sampling or “Other Reasonable Methods”

- The preamble of the proposed rules places considerable emphasis on statistical sampling (several pages), but provides virtually no discussion of “other reasonable methods”
 - Brief mention of “statistical inference or numerically solved median estimate,” but only in the context of conducting a stratified sampling approach
- Statistical sampling may be a viable approach for a number of companies but for others, it might not be as cheap and easy as the SEC implies
 - Extracting sample data from multiple payroll systems (e.g., manually, computer program, etc.)
 - Determination of sample size and confidence intervals
 - Effort spent sampling might exceed effort spent arraying data and finding median directly
 - Sampling requires ability to access all of the data anyway
 - Data for certain units or segments may not be available (e.g., non-US unit run independently) which makes sampling impossible
 - But, could make inferences about variance if descriptive statistics provided for unit
- Other reasonable methods might include:
 - Statistical inference
 - SEC cites paper that discusses the use of weighted-average medians, but only in the context of stratified cluster sampling
 - Simulation
 - Not mentioned proposed rules

Simulation Methodology

- Proposed simulation assumes compensation data follow a “lognormal distribution”
 - Common assumption with academic and BLS research (including SEC citations in proposed rules)
 - Data can't be below \$0 and are assumed to be skewed to the right with mean higher than median
 - Adjust compensation data by taking natural log (i.e., log base e or LN(x) in Excel)
 - Ranking of data is exactly the same before and after data are adjusted
 - Adjusted data assumed to follow a normal distribution (bell curve)
 - Normal distribution has desirable mathematical properties (e.g., symmetrical with mean = median)
 - Properties of normal distribution are well-known and easily applied to create simulation approach
 - Other distributions would be more complicated
- Once a normal distribution can be assumed, simulation only requires:
 - Mean = μ
 - Standard deviation = σ
 - Stochastic factor from random number generator (e.g., NORM.S.INV(RAND()) in Excel) = z

Simulation Methodology (continued)

- If quartile data are provided for a business unit, then can take the following steps:
 1. Take the natural log of each quartile
 - These data are now assumed to be normally distributed with mean = median
 2. Log-adjusted median data is also mean (μ) of normal distribution
 3. Algebraically solve for the standard deviation
 - Since log-adjusted data are assumed to be normally distributed (i.e., symmetrical bell-curve), the statistical properties of a normal distribution can be applied
 - These properties are well-known and can be found in virtually all statistics text books
 - 25th percentile is -0.675 standard deviations below the mean (median) and 75th percentile is +0.675 standard deviation above the mean (median)
 - Solve for standard deviation from these quartiles
 - Take average for standard deviation input (σ)
 - Can use other percentiles if available (e.g., the 16th and 84th percentiles are ± 1 standard deviation from mean)
 4. Come up with a random number between 0 and 1 (e.g., RAND() function in Excel) and convert to a standard normal random variable (z), (e.g., use table from textbook or NORM.S.INV in Excel)
 5. Simulate log-adjusted compensation = $\mu + z \sigma$
 6. Repeat for the number of remaining employees in the business unit
- Follow the above process for each business unit where data are lacking
 - Approach implicitly assumes no correlation between different units
- Combine all actual and simulated data into one series and take median
- Take the anti-log (i.e., e^x) of the result to find median pay (and other percentiles)

Simulation Example

- Quartile data were provided for Global population as well as US and non-US populations
- Test approach by trying to replicate median of global data by using only US and non-US quartile data
- Data provided below (actual and log-adjusted):

Salary (\$ US)			
	<i>Global</i>	<i>Non-US</i>	<i>US</i>
<i>Percentile</i>	<i>n = 50,000</i>	<i>n = 31,000</i>	<i>n = 19,000</i>
25th	\$ 11,250	\$ 5,750	\$ 35,750
50th	\$ 34,250	\$ 17,250	\$ 51,500
75th	\$ 56,500	\$ 41,000	\$ 77,750

ln(Salary)			
	<i>Global</i>	<i>Non-US</i>	<i>US</i>
<i>Percentile</i>	<i>n = 50,000</i>	<i>n = 31,000</i>	<i>n = 19,000</i>
25th	9.33	8.66	10.48
50th	10.44	9.76	10.85
75th	10.94	10.62	11.26

- Mean for Non-US is 9.76 and mean for US is 10.85
- Solve for standard deviations:
 - Non-US: $[(8.66 - 9.76) \div -0.675 + (10.62 - 9.76) \div 0.675] \div 2 = [1.63 + 1.27] \div 2 = 1.45$
 - US: $[(10.48 - 10.85) \div -0.675 + (11.26 - 10.85) \div 0.675] \div 2 = [0.55 + 0.61] \div 2 = 0.58^*$
- Use Excel RAND() and NORM.S.INV functions and simulate data point
 - Example: $\text{NORM.S.INV}(\text{RAND}()) = 0.4232 \rightarrow$ for US data point, $10.85 + 0.4232 \times 0.58 = 11.10$;
 $e^{11.10} = \$66,171$
- Simulate 31,000 data points for non-US and 19,000 for US, combine into one series and determine median; repeat many times and average all results

* Note: Might be more appropriate to take average of variance rather than average of standard deviations; difference is negligible in this example

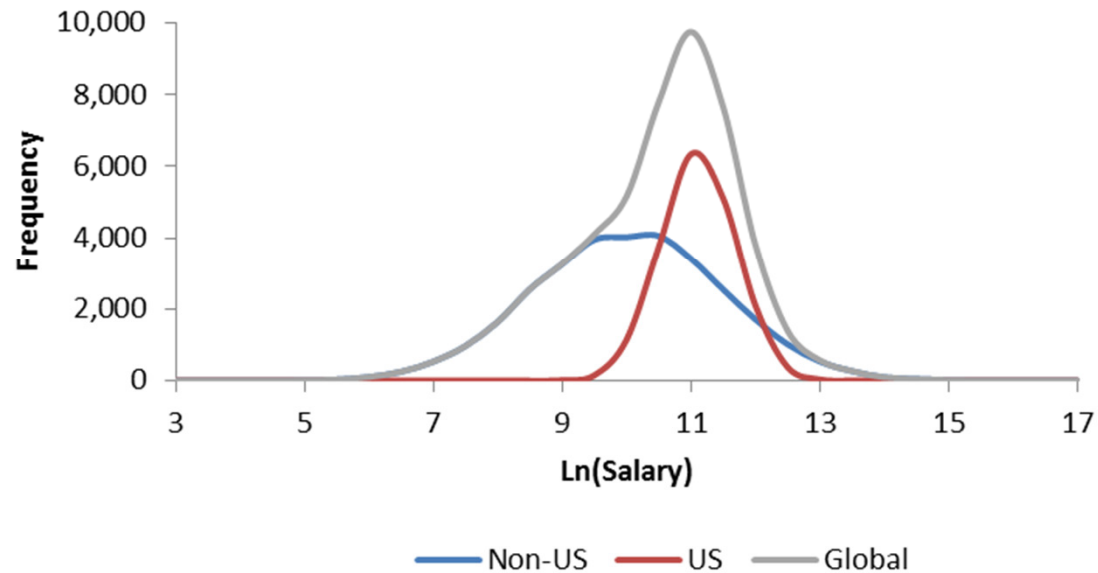
Simulation Example (continued)

- Results from one simulation is provided as follows:

Ln(Salary)			
	<i>Global</i>	<i>Non-US</i>	<i>US</i>
<i>Percentile</i>	<i>n = 50,000</i>	<i>n = 31,000</i>	<i>n = 19,000</i>
25th	9.39	8.77	10.47
50th	10.42	9.77	10.86
75th	11.07	10.76	11.23

Salary (\$ US)			
	<i>Global</i>	<i>Non-US</i>	<i>US</i>
<i>Percentile</i>	<i>n = 50,000</i>	<i>n = 31,000</i>	<i>n = 19,000</i>
25th	\$ 12,009	\$ 6,428	\$ 35,286
50th	\$ 33,460	\$ 17,544	\$ 51,807
75th	\$ 64,283	\$ 46,906	\$ 75,592

Distribution of Simulated Ln(Salary)



Simulation Example (continued)

- The following table compares the results from this one simulation to actual (the difference between the two are displayed):

Salary (\$ US)			
	<i>Global</i>	<i>Non-US</i>	<i>US</i>
<i>Percentile</i>	<i>n = 50,000</i>	<i>n = 31,000</i>	<i>n = 19,000</i>
25th	\$ 759	\$ 678	\$ (464)
50th	\$ (790)	\$ 294	\$ 307
75th	\$ 7,783	\$ 5,906	\$ (2,158)

- Note that this was just one simulation
 - The process can be repeated many (e.g., 10,000) times
 - Take an average of the results to get a more robust answer

Issues to Consider

- This approach could be useful when trying to integrate multiple data series
 - Dispersion for a business unit could influence position (ranking) of data in the global population
 - Actual data are less important than the ranking when finding a median
 - Unlike a mean, magnitude of data doesn't influence the middle-point
- Approach heavily relies on assumption that pay data follow a lognormal distribution
 - Actual distribution might differ (e.g., note that quartiles in example showed some skewness)
 - SEC proposal cites BLS and academic studies that rely on this assumption
 - Proposed rules do not appear to prohibit reasonable assumptions like this, but it does criticize variances that are “generally assumed”
 - Using descriptive statistics that are specific to a company or unit addresses this concern
 - Specific guidance from the SEC would help to quell any potential criticism
 - Simulation based on other distributions might be developed (if these can be identified)
 - Less practical (more complicated) – need to incorporate skewness, kurtosis into model
- The example shows evidence that simulation would be viable, but not proof
 - The answer (i.e., the overall median) is already known
- The approach relies on obtaining some measure(s) of dispersion around the median
 - If only median data are provided, then deriving input for the standard deviation would be impossible
 - Simulation approach would need to rely on additional assumptions
 - Other approaches may be more palatable
- Simulation demonstrated does not consider correlation between different groups of data
 - Implicit assumption is that there is no correlation between data sets
 - Can incorporate correlation into simulation

Recommendations to the SEC

- If the SEC goes forward with issuing final rules, hopefully it will consider the following:
 - Alternatives for determining pay ratio should focus on practicality and cost reduction, not on spurious precision
 - Inherent tradeoff between incremental cost and incremental benefit of more “precision” (e.g., would an investor’s actions change if the ratio were higher/lower by 50?)
 - Explicitly permit (1) use of lognormal assumption and (2) use of formulaic or numerical approaches to estimate median pay (e.g., simulation)
 - On the one hand, proposed rules rely heavily on studies that make lognormal assumption (footnote 204), but criticizes the assumption elsewhere (footnote 83 – but only in the context of determining sample size)
 - Using descriptive statistics to approximate company- or segment-specific compensation variance should address this criticism
 - Allow companies to provide a reasonable range of ratios based on reasonable assumptions
 - For example, if using salary to identify “median employee:”
 - Assuming \$0 for all other elements would provide lowest value (and highest, most conservative ratio)
 - Assuming “median employee’s” age and service qualifies him for early retirement provides highest pension change
 - If a range of outcomes is deemed acceptable, then any effort spent being more “accurate” would be a waste of shareholders money
- Full comment letter sent to SEC: <http://www.sec.gov/comments/s7-07-13/s70713-570.pdf>