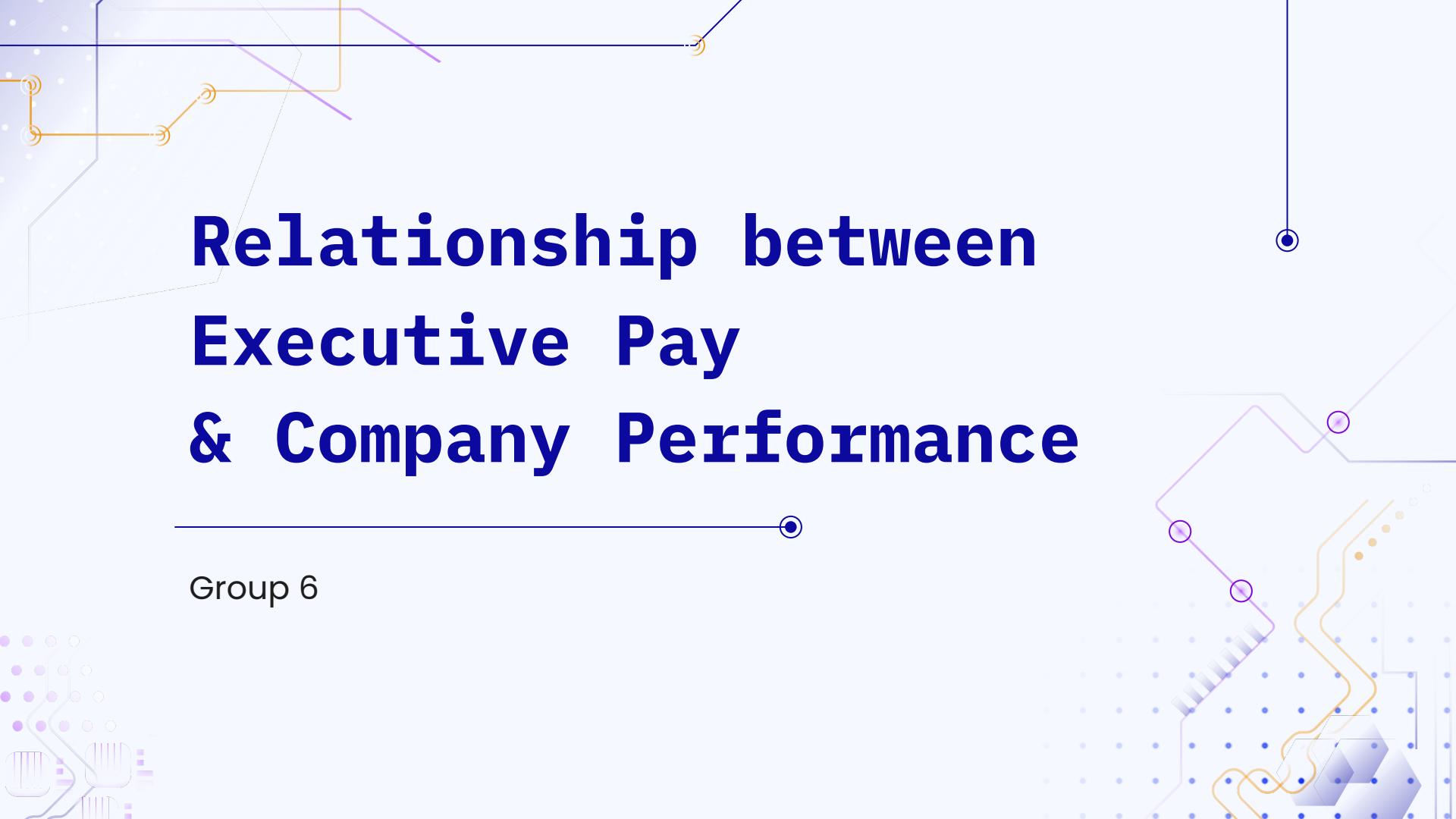


# **Relationship between Executive Pay & Company Performance**

Group 6





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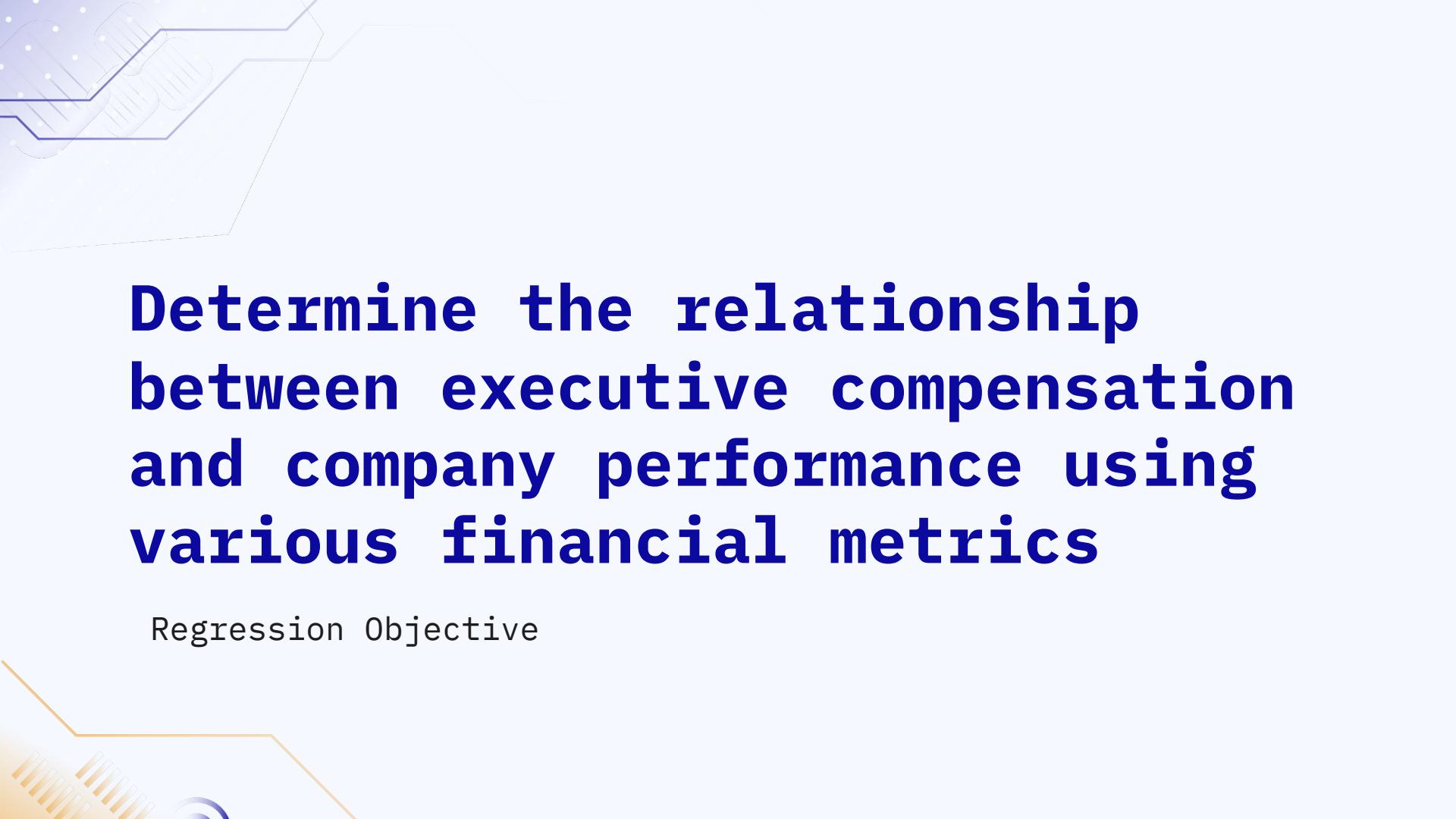
**06** Conclusion

01

# Introduction

---





**Determine the relationship  
between executive compensation  
and company performance using  
various financial metrics**

Regression Objective

# Literature Review

In developed cities such as Australia, Japan and Singapore, both the CEO's compensation and the total salaries were higher when a company did better.(Kayani, Umar & Christopher, 2022).

Higher CEO pay is seen to be a great motivator to CEOs to perform better. (Kweh et al., 2022)

## Central Hypothesis

Higher CEO compensation will lead to better firm performance

→ Analyse the relationship between CEO Compensation and company's financial performance & determine which compensation components affect financial performance significantly

# Dataset Overview

## Dataset Extraction

All Datasets were extracted from Wharton Research Data Services under Compustat – Capital IQ. Two datasets were extracted: Company Financial Data and Executive Compensation Data

### Company Financial Data

```
• comp_data      261959 obs. of 41 variables
  $ gvkey    : chr [1:261959] "001004" "001004" "001004" "001004" ...
  $ datadate: Date[1:261959], format: "2000-05-31" "2001-05-31" "2002-...
  $ fyyear   : num [1:261959] 1999 2000 2001 2002 2003 ...
  $ infdfmt  : chr [1:261959] "INDL" "INDL" "INDL" "INDL" ...
  $ consol   : chr [1:261959] "C" "C" "C" "C" ...
  $ popsrc   : chr [1:261959] "D" "D" "D" "D" ...
  $ datafmt  : chr [1:261959] "STD" "STD" "STD" "STD" ...
  $ cusip    : chr [1:261959] "000361105" "000361105" "000361105" "0003...
  $ conn     : chr [1:261959] "AAR CORP" "AAR CORP" "AAR CORP" "AAR COR...
  $ curcd   : chr [1:261959] "USD" "USD" "USD" "USD" ...
  $ fyr      : num [1:261959] 5 5 5 5 5 5 5 5 5 ...
  $ act      : num [1:261959] 511 486 437 396 432
```

### Executive Compensation Data

```
• exec_data      44983 obs. of 22 variables
  $ EXEC_FULLNAME   : chr "David P. Storch" "David P. Storch" ...
  $ CO_PER_ROL      : int 5623 5623 5623 5623 5623 5623 5623 5623 ...
  $ CONAME          : chr "AAR CORP" "AAR CORP" "AAR CORP" "AA...
  $ CEOANN          : chr "CEO" "CEO" "CEO" "CEO" ...
  $ SALARY          : num 676 665 661 661 695 ...
  $ BONUS           : num 270 0 0 496 592 ...
  $ OTHCOMP         : num 139 101 151 152 146 ...
  $ TOTAL_CURR      : num 946 665 661 1158 1287 ...
  $ PENSION_VALUE_TOT: num NA NA NA NA NA ...
  $ OTHANN          : num 0 0 62 67.8 69.5 ...
  $ RSTKGRNT        : num 0 0 0 348 695 ...
  $ OPTTQN_AWARDS_RISK_VALUE: num 2825 1281 1268 2668 4917
```

# 02

# Data Cleaning and Preparation

---

# Data Set Cleaning

## Company Financial Data

### R-script

```
#fill in fyear if missing, based on Compustat's May cutoff
comp_data$fyear <- ifelse(
  is.na(comp_data$fyear),
  ifelse(
    as.numeric(format(comp_data$datadate, format = "%m")) > 5,
    as.numeric(format(comp_data$datadate, format = "%Y")),
    as.numeric(format(comp_data$datadate, format = "%Y")) - 1), comp_data$fyear)

# filtering only "INDL" data
comp_data <- filter(comp_data, indfmt == "INDL")

# creating firm-year indices using gvkey and fyear
comp_data <- arrange(comp_data, gvkey, fyear)
comp_data$index <- paste(comp_data$gvkey, comp_data$fyear, sep = "_")

#2.4 identifying and removing duplicate indices
length(unique(comp_data$index)) #291933 different from the base data set

comp_data_clean <- subset(
  comp_data,
  !(index %in% subset(comp_data, duplicated(index) == 1)$index)
)
nrow(comp_data_clean) #261907 rows
length(unique(comp_data_clean$index)) #261907 firm-year indices
```

### Functions

#### 1. Replacing missing NA values in *fyear*

NA values caused by May month cutoff in Compustat database. If *datadate* > 5, *fyear* remain, otherwise *fyear* to be replaced by the previous year.

#### 2. Filtering based on industries in *indfmt*

Original dataset consists of companies both industrial and financial in nature. Filter to include only industrial companies.

#### 3. Creating *index* and removing duplicated *index*

*index* variable is created, combining *gvkey* and *fyear*.

Duplicated rows caused by multiple observation in the same *fyear*. hence are removed to have a uniquely identified observation by the *index* variable.

# Dataset Cleaning

## Executive compensation data

### R-script

```
#4.1 Filtering to only including the data to CEO
exec_data <- exec_data[exec_data$CEOANN == "CEO", ]  
  
#4.2 converting GVKEY to character
exec_data$GVKEY <- as.character(exec_data$GVKEY)
exec_data$GVKEY <- ifelse(nchar(exec_data$GVKEY) == 4, paste0("00", exec_data$GVKEY), exec_data$GVKEY)
exec_data$GVKEY <- ifelse(nchar(exec_data$GVKEY) == 5, paste0("0", exec_data$GVKEY), exec_data$GVKEY)  
  
#4.4 creating firm-year indices using gvkey and year
exec_data <- arrange(exec_data, GVKEY, YEAR, CO_PER_ROL)
exec_data$index <- paste(exec_data$GVKEY, exec_data$YEAR, sep = "_")  
  
#4.5 identifying and removing duplicate indices
length(unique(exec_data$index)) #44940 different from the base data set  
  
exec_data_clean <- subset(
  exec_data,
  !(index %in% subset(exec_data, duplicated(index) == 1)$index)
)
nrow(exec_data_clean) #44897 rows
length(unique(exec_data_clean$index)) #44897 firm-year indices
```

### Functions

#### 1. Filtering by “CEO” in CEOANN

CEOANN indicates the executive that served as CEO for all or most of fiscal year.

Filter to shortlist the CEO compensation

#### 2. Converting GVKEY to a character variable

After conversion, 4 or 5 digits under GVKEY should be added with 2 or 1 zeros in front of the characters. Convert to provide ease of merging with the company dataset.

#### 3. Creating index and removing duplicated index

Same steps followed in company financial data

# Combining the datasets



## R-script

```
data_comb <- inner_join(comp_data_clean, exec_data_clean, by = "index")
```

evt	sale	seq	xint	index	EXEC_FULLNAME	CO_PER_ROL	CONAME
874.3	874.3	340.2	21.88	001004_2000	David P. Storch	5623	AAR CORP
638.7	638.7	310.2	19.79	001004_2001	David P. Storch	5623	AAR CORP
606.3	606.3	295.0	19.53	001004_2002	David P. Storch	5623	AAR CORP
652.0	652.0	301.7	18.81	001004_2003	David P. Storch	5623	AAR CORP
747.8	747.8	314.7	16.41	001004_2004	David P. Storch	5623	AAR CORP

- `inner join()`:

Merge 2 datasets based on common columns and returns only the rows with matching keys in both data frames.

- **by = “index” :**

The combination of `gvkey` and `fyear`, a unique key to simplify merging process and avoid duplicates in the merged data frames.

## Output

 data_comb	44535 obs. of 63 variables
\$ gvkey	: chr [1:44535] "001004" "001004"
\$ datadate	: Date[1:44535], format: "2001-05
\$ fyyear	: num [1:44535] 2000 2001 2002 20
\$ indfmt	: chr [1:44535] "INDL" "INDL" "IN
\$ consol	: chr [1:44535] "C" "C" "C" "C" .
\$ popsrc	: chr [1:44535] "D" "D" "D" "D" .
\$ datafmt	: chr [1:44535] "STD" "STD" "STD"
\$ cusip	: chr [1:44535] "000361105" "0003
\$ comm	: chr [1:44535] "AAR CORP" "AAR C
\$ curcd	: chr [1:44535] "USD" "USD" "USD"
\$ fyr	: num [1:44535] 5 5 5 5 5 5 5 5 5 5

- **exec\_data\_clean**: 44897 obs.
  - **comp\_data\_clean**: 261959 obs.
  - **data\_comb**: 44535 obs.

There are 44535 rows with matching keys ->  
A more focused dataset for regression  
modeling

# Calculating Ratios



## 1. Generate lagged values

A prerequisite to compute ratios such as return on assets, return on equity and inventory turnover, which require lagged values of total asset *at*, inventory *invt*, and total *equity seq* to compute averages.

### R-script

```
data_comb_2 <- arrange(data_comb_1, index)
data_comb_2 <- data_comb_1 %>% group_by(gvkey) %>% mutate(at_lag = ifelse(fyear == lag(fyear) + 1, lag(at, n = 1), NA),
                                                     invt_lag = ifelse(fyear == lag(fyear) + 1, lag(invt, n = 1), NA),
                                                     seq_lag = ifelse(fyear == lag(fyear) + 1, lag(seq, n = 1), NA)) %>% ungroup()
```

### Functions

- **arrange()**: Sort the observations in ascending order based on *index* variable
- **group\_by()**: Group the data by the *gvkey* variable
- **mutate()**: Create lagged variables (*at\_lag*: Lagged total assets for the previous year)
- **Ifelse()**: Apply condition that lagging is applied only if the current year in *fyear* is one year after the lagged year, and if the condition is not met, NA is assigned.

### Output

index	gvkey	at	at_lag	invt	invt_lag	seq	seq_lag
1	001004_2000	001004	701.9	NA	320.59	NA	340.21
2	001004_2001	001004	710.2	701.9	286.59	320.59	310.24
3	001004_2002	001004	686.6	710.2	259.95	286.59	294.99
4	001004_2003	001004	709.3	686.6	247.25	259.95	301.68
5	001004_2004	001004	732.2	709.3	255.48	247.25	314.74
6	001004_2005	001004	978.8	732.2	323.59	255.48	422.72
7	001004_2006	001004	1067.6	978.8	342.59	323.59	494.24
8	001004_2007	001004	1362.0	1067.6	435.61	342.59	585.25
							494.2

▶ data_comb_1	44535 obs. of 63 variables
▶ data_comb_2	44535 obs. of 66 variables

# Calculating Ratios



## 2. Generate ratio variables

Selected ratios from 4 distinct categories acts as the measure of firm financial performance and will be utilized as the dependent variables/ controlled variables in our regression analysis.

### R-script

```
data_comb_3 <- data_comb_2 %>% mutate(net_profit_margin=ni/revt,operating_profit=oiadp/revt,ROE=ni/((seq+seq_lag)/2),  
current_ratio=act/lct,cash_ratio=ch/lct,debt_ratio=lt/at,debt_to_equity_ratio=lt/seq,  
asset_turnover=revt/(at-lt),roa=ni/((at+at_lag)/2),inventory_turnover=cogs/((invt+invt_lag)/2))
```

### Functions

- **mutate()**: Add ratio variables to the dataset based on calculations. 10 key financial ratio are calculated for potential integration into regression analysis and predictive modeling in r.
- **Result:** The dataset 'data\_comb\_3' has been enriched with the financial ratios. Further data cleaning is essential to address outliers and handle NA/inf values.

### Output

```
$ net_profit_margin      : num [1:44535] 0.0212 -0.0  
$ operating_profit       : num [1:44535] 0.05238 0.0  
$ ROE                   : num [1:44535] NA -0.1812  
$ current_ratio          : num [1:44535] 3.87 2.9 1.  
$ cash_ratio              : num [1:44535] 0.11 0.229  
$ debt_ratio              : num [1:44535] 0.515 0.563  
$ debt_to_equity_ratio    : num [1:44535] 1.06 1.29 1  
$ asset_turnover          : num [1:44535] 2.57 2.06 2  
$ roa                     : num [1:44535] NA -0.08348  
$ inventory_turnover      : num [1:44535] NA 1.73 1.8
```

# 03

# Sample Period and Restrictions

---



# 2000-2022: 22 years



## Long-term Trends and patterns

Firm or market  
specific factors



## Changes in the economy, regulation and society

New regulations, laws  
and societal changes



## Global Events

Financial crisis,  
pandemics,  
international conflicts

# 6 compensation components

## 1. Salary

- Identification of base pay
- Wage growth

## 2. Bonus

- Usually with better firm performance, the bonuses increase in value.
- Analysis over time uncovers trends in performance-based pay models.

## 3. Other Compensation (OTHCOMP)

- Other benefits, perks and non-standard compensation items

## 4. Restricted Stock Grants (RSTKGRNT)

- Equity-based compensation scheme
- To align CEO's interests with the performance of a company by offering a stake in the company

## 5. Option Awards Block Value (OPTION\_AWARDS\_BLK\_VALUE):

- How companies use options as a tool to incentivise and motivate the CEOs

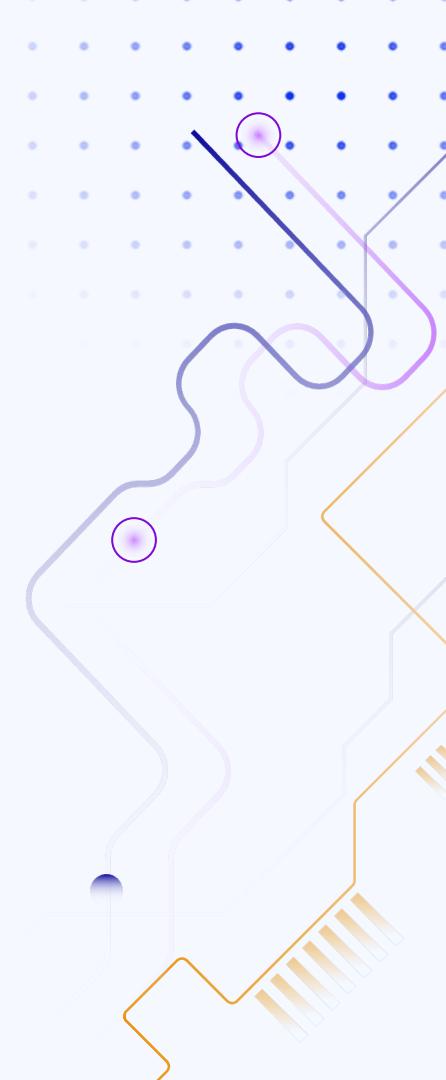
## 6: Long-Term Incentive Plan (LTIP)

- LTIPs are often designed with the objective of achieving a company's long-term objectives and retaining key management, such as the CEO.
- whether a company is able to adapt to changing business environments and maintain financial performance

# 04

# Statistic Model

---



# Statistic Model

## Aim of Statistic Model

Analyse the relationship between CEO Compensation and company's financial performance & determine which compensation components affect financial performance significantly

$$\text{Financial Ratio} = \beta_0 + \beta_1 \times \text{Comp Component} + \text{Control Variables}$$

- IV - Compensation Components:**

Salary, Bonus, Other Compensation, Restricted Stock Grant, Stock Options Awarded, Long-term Incentive Pay

- DV - Financial Ratios:**

Profitability (Net Profit Margin), Liquidity (Current Ratio), Leverage (Debt to Equity Ratio), Operational Efficiency (Return on Assets)

# Statistic Model

Financial Ratio	Control Variable
<b>Profitability</b> - Profit Margin	<i>at</i>
<b>Efficiency</b> - Return on Assets	<i>debt_to_equity_ratio</i> <i>ceo_years</i>
<b>Liquidity</b> - Current Ratio	<i>at</i> <i>debt_to_equity_ratio</i> <i>asset_turnover</i>
<b>Leverage</b> - Debt to Equity Ratio	<i>at</i> <i>net_profit_margin</i> <i>asset_turnover</i>

## Purpose

- ***at (Total Assets)***: Account for the size of the firm
- ***debt\_to\_equity\_ratio***: Account for the variations in firm's capital structure
- ***ceo\_years***: Account for variations in leadership experience
- ***Asset\_turnover***: Account for variations in operational efficiency of the firm
- ***net\_profit\_margin***: Account for variations in profitability and financial health of the firm

# Statistic Model

## Further data cleaning

### R-script

```
data_reg_3a <- data_reg_3 %>%
  filter(!is.na(at) & !is.infinite(inventory_turnover))

#7.3 remove outliers with truncation method
ggplot(data_reg_3a, aes(x = fyear, y = net_profit_margin)) + geom_point() + geom_smooth(method = "lm", se = FALSE)
ggplot(data_reg_3a, aes(x = fyear, y = debt_to_equity_ratio)) + geom_point() + geom_smooth(method = "lm", se = FALSE)

data_reg_3a %>% filter(!net_profit_margin > quantile(net_profit_margin, 0.99, na.rm = TRUE)
  & !net_profit_margin < quantile(net_profit_margin, 0.01, na.rm = TRUE))
```

### Functions

#### 1. Remove NA and inf values under ratio variables

Filter out NA values in total asset as it is unlikely to be 0 for a company.

Remove inf values in the ratios to prevent error during regression

#### 2. Analyse and remove outliers

Analyse the outliers by plotting *fyear* (x-axis) against various ratios to identify presence of outliers.

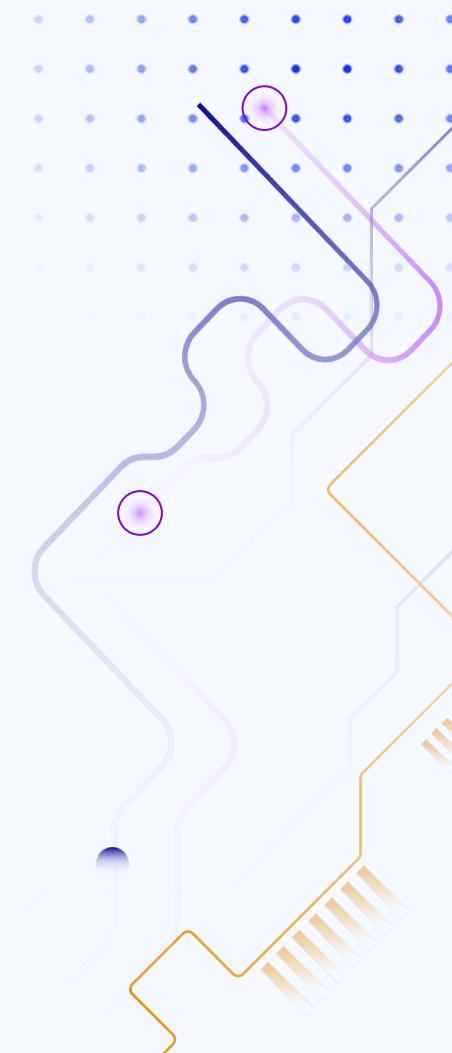
Filter out top and bottom 0.01 quantile to truncate potential outliers affecting regression model.

Limit CEO years to 20 due to a notable decrease in density beyond that threshold on the plot

# 05

# Regressions

---



# Regression models

## Regressing on training dataset

```
#8.1. choosing training and test dataset  
set.seed(1)  
train1 <- sample_frac(data_reg_3d_clean, 0.75)  
test1 <- anti_join(data_reg_3d_clean, train1)
```

### Model 1 (IV + DV + CV)

```
lm1a_train <-  
lm(net_profit_margin ~ salary_at + at  
+ debt_to_equity_ratio + ceo_years,  
train1)
```

### Model 2 (stepwise, forward and backward)

1. step(lm1a\_train, direction = "both")
2. step(lm1a\_train, direction = "forward")
3. step(lm1a\_train, direction = "backward")

### Model 3 (IV + DV + FE)

```
data_reg_fixeff_1a <-  
felm(net_profit_margin ~ salary_at +  
at + debt_to_equity_ratio +  
ceo_years | gvkey + fyear | 0 | gvkey  
+ fyear, train1)
```

# Salary vs Firm Financial Performance

IV: Salary/Total asset, DV: Net profit margin		IV: Salary/Total asset, DV: Return on asset	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0348	Adj. R squared = 0.329	Adj. R squared = 0.0379	Adj. R squared = 0.329
RSE = 0.139	RSE = 0.116	RSE = 0.0882	RSE = 0.0701
Salary_at (slope) = -0.0119	Salary_at (slope) = -0.00384	Salary_at (slope) = -0.00815	Salary_at (slope) = -0.00410
Salary_at (P value) < 0.1	Salary_at (P value) > 0.1	Salary_at (P value) < 0.1	Salary_at (P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = -19084		AIC = -34960	
RMSE = 0.1386		RMSE = 0.08696	

- Extremely low adj. R squared -> Limited explanatory power for profit margin.
- With fixed effects: Improvement in adj. R squared and RSE. P-value rises (>0.1), signaling overfitting .
- Negative slope indicates a negative relationship

- Consistent P-values < 0.1, indicating statistical significance
- R-squared with fixed effects increases to 0.329 (still relatively low)
- Negative relationship with return on assets, effect size remains low, similar to the previous regression.

# Salary vs Firm Financial Performance

IV: Salary/Total asset, DV: Current Ratio	
Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.108	Adj. R squared = 0.719
RSE = 1.41	RSE = 0.794
Salary_at (slope) = 0.119	Salary_at (slope) = -0.04943
Salary_at (P value) < 0.1	Salary_at (P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 61129	
RMSE = 1.436	

- Despite statistical significant of Model 1&2, adj. R squared is extremely low.
- With fixed effects, R squared hike and there is dynamic change where negative relationship emerges.
- Returns highest AIC compared to previous regressions.

IV: Salary/Total asset, DV: Debt to Equity	
Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.468	Adj. R squared = 0.795
RSE = 1.88	RSE = 1.16
Salary_at (slope) = -0.1169	Salary_at (slope) = -0.0393
Salary_at (P value) < 0.1	Salary_at (P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 70952	
RMSE = 1.983	

- Model 3: Improved adj. R-squared, lower RSE and IV-DV relationship remain statistically significant.
- A reduced influenced of salary on DE ratio with a slope of -0.0393
- Negative slope indicates an inverse relationship

# Salary vs Firm Financial Performance

## Summary

- Slope coefficient close to 0 shows a weak association between salary and firm financial performance.
- Extremely low adjusted r-squared shows a limited model ability to explain profit margin and ROA variability.
- Regression with Debt to equity ratio: P-value < 0.1 shows a statistically significant relationship, and 79.5% of the variability in the DV is explained.

## Model 3 with Fixed effects

$$\begin{aligned}\text{Debt-to-equity-ratio} = & -0.039374 * \text{salary\_at} + 0.00000399 * \text{total asset} \\ & - 0.63801114 * \text{profit margin} + 0.76662036 * \text{asset turnover}\end{aligned}$$

# Bonus vs Firm Financial Performance

IV: Bonus/Total asset, DV: Net Profit Margin		IV: Bonus/Total asset, DV: Return on asset	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0154	Adj. R squared = 0.33	Adj. R squared = 0.0143	Adj. R squared = 0.389
RSE = 0.141	RSE = 0.116	RSE = 0.0893	RSE = 0.0703
bonus_at (slope) = -0.00595	bonus_at (slope) = 0.00825	bonus_at (slope) = -0.00212	bonus_at (slope) = -0.00477
bonus_at (P value) < 0.1	bonus_at (P value) < 0.1	bonus_at (P value) < 0.1	bonus_at (P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = -18739		AIC = 70952	
RMSE = 0.1393		RMSE = 1.983	

- Negative slope coefficient is statistically significant
- Firm and year fixed effect improved adj. r-squared, though remains low
- IV coefficient shifted to positive in Model 3

- In alignment with previous regression, adj. r squared remains relatively low across models
- Model with Fixed effects cause elevated p-value, suggesting potential overfitting and unreliable estimates

# Bonus vs Firm Financial Performance

IV: Bonus/Total asset, DV: Current Ratio		IV: Bonus/Total asset, DV: Debt to Equity Ratio	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0945	Adj. R squared = 0.717	Adj. R squared = 0.464	Adj. R squared = 0.795
RSE = 1.42	RSE = 0.795	RSE = 1.88	RSE = 1.17
bonus_at (slope) = 0.0979	bonus_at (slope) = -0.0398	bonus_at (slope) = -0.1621	bonus_at (slope) = -0.02845
bonus_at (P value) < 0.1	bonus_at (P value) < 0.1	bonus_at (P value) < 0.1	bonus_at (P value) > 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 61396		AIC = 71081	
RMSE = 1.449		RMSE = 1.995	

- Substantial improvement in explanatory power and precision
- Outperforms previous models with profit margin and ROA
- P-value below threshold: Statistically significant negative relationship between CEO bonus and current ratio.

- Model 1 & 2 with moderate adj. r-squared with p-value < 0.1
- With fixed effects, explanatory power of model is improved but p-value above threshold, potential lack of statistical significance

# Bonus vs Firm Financial Performance

## Summary

- Very weak relationship indicated by exceedingly low adjusted r-squared and negative coefficient close to 0
- When regressing with debt-to-equity ratio, incorporation of fixed effects results in increase of p-value above threshold, raising concern about model robustness
- With current ratio as DV, regression findings provide viable basis for investigating its association with CEO bonus

## Model 3 with Fixed effects

$$\begin{aligned}\text{Current ratio} = & -0.03978 * \text{bonus\_at} - 0.000001211 * \text{total asset} \\ & + 0.005079 * \text{debt to equity ratio} - 0.03595 * \text{asset turnover}\end{aligned}$$

# Other comp vs Financial Performance

IV: Other compensation/Total asset, DV: Net profit margin		IV: Other compensation/Total asset, DV: Return on asset	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0183	Adj. R squared = 0.329	Adj. R squared = 0.0199	Adj. R squared = 0.391
RSE = 0.1406	RSE = 0.116	RSE = 0.0890	RSE = 0.0702
othcomp_at (slope) = -0.0082	othcomp_at (slope) = -0.00417	othcomp_at (slope) = -0.0064	othcomp_at (slope) = -0.003293
othcomp_at (P value) < 0.1	othcomp_at (P value) < 0.1	othcomp_at (P value) < 0.1	othcomp_at (P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = -67951		AIC = -83798	
RMSE = 0.1387		RMSE = 0.087	

- Extremely low adj. R squared -> Limited explanatory power for profit margin.
- With fixed effects: Improvement in adj. R squared and RSE indicating better explanation in variation of profit margin.
- Negative slope indicates a negative relationship

- Consistent P-values < 0.1, indicating statistical significance
- R-squared with fixed effects increases to 0.329 (still relatively low)
- Negative relationship with return on assets, effect size remains low, similar to the previous regression.

# Other comp vs Financial Performance

IV: Other compensation/Total asset, DV: Current Ratio		IV: Other compensation/Total asset, DV: Debt to Equity Ratio	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0932	Adj. R squared = 0.717	Adj. R squared = 0.4622	Adj. R squared = 0.794
RSE = 1.42	RSE = 0.796	RSE = 1.88	RSE = 1.17
othcomp_at (slope) = 0.0404	othcomp_at (slope) = -0.000506	othcomp_at (slope) = -0.0514	othcomp_at (slope) = -0.02091
othcomp_at (P value) < 0.1	othcomp_at (P value) > 0.1	othcomp_at (P value) < 0.1	othcomp_at (P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 12261		AIC = 21,970	
RMSE = 1.448		RMSE = 1.994	

- Substantial improvement in explanatory power and precision
- Outperforms previous models with profit margin and ROA
- P-value above threshold 0.1: Statistically insignificant negative relationship between CEO bonus and current ratio.

- Model 1 & 2 with moderate adj. r-squared with p-value < 0.1
- With fixed effects, explanatory power of model is significantly improved (BEST MODEL)

# Other comp vs Financial Performance

## Summary

- Weak relationship indicated by exceedingly low adjusted r-squared and negative coefficient close to 0
- When regressing with debt-to-equity ratio, incorporation of fixed effects results in increase of adjusted R square to 0.794 without raising the p-value about 0.1 threshold
- With debt-to-equity ratio as DV, regression findings provide viable basis for investigating its association with CEO's other compensation

## Model 3 with Fixed effects

$$\text{Debt to equity-ratio} = -0.02106 * \text{othcomp\_at} + 0.00000394 * \text{total asset} - 0.6299 * \text{profit margin} + 0.76648 * \text{asset turnover}$$

# RSTKGRNT vs Firm Financial Performance

IV: RSTKGRNT, DV: Net profit margin		IV: RSTKGRNT, DV: Return on asset	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0147	Adj. R squared = 0.329	Adj. R squared = 0.00517	Adj. R squared = 0.388
RSE = 0.141	RSE = 0.116	RSE = 0.0897	RSE = 0.0703
RSTKGRNT (slope) = 0.000002162	RSTKGRNT (slope) = -0.000001928	RSTKGRNT (slope) = 0.000002254	RSTKGRNT (slope) = 0.000001065
RSTKGRNT (P value) < 0.1	RSTKGRNT (P value) < 0.1	RSTKGRNT (P value) < 0.1	RSTKGRNT (P value) > 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = -18736		AIC = -34387	
RMSE = 0.1391		RMSE = 0.08773	

- Extremely low adj. R squared -> Limited explanatory power for profit margin.
- With fixed effects: Improvement in adj. R squared.
- Slope close to 0 indicates a weak or no relationship

- P-value rises (>0.1), signaling overfitting .
- R-squared with fixed effects increases(still relatively low)
- Near 0 correlation = effect size remains low

# RSTKGRNT vs Firm Financial Performance

IV: RSTKGRNT, DV: Current Ratio	
Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0936	Adj. R squared = 0.717
RSE = 1.42	RSE = 0.796
RSTKGRNT (slope) = -0.000051612	RSTKGRNT (slope) = 0.000001804
RSTKGRNT (P value) < 0.1	RSTKGRNT (P value) > 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 61406	
RMSE = 1.449	

- Improvement in r square when fixed effects are added
- P-value rises (>0.1), signaling overfitting .
- Negative correlation improves, but still a near 0 figure.

IV: RSTKGRNT, DV: Debt to Equity	
Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.462	Adj. R squared = 0.795
RSE = 1.89	RSE = 1.17
RSTKGRNT (slope) = - 0.000017647	RSTKGRNT (slope) = -0.00000966
RSTKGRNT (P value) > 0.1	RSTKGRNT (P value) > 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 71135	
RMSE = 1.995	

- Rise in value of r square, with highest value of 0.795 being observed
- This model has returned the highest AIC, indicating a weaker relationship fit between RSTKGRNT and the debt-to-equity ratio
- Negative slope indicates an inverse relationship

# RSTKGRNT vs Firm Financial Performance

## Summary

- Slope coefficient close to 0 shows a weak association between RSTKGRNT and firm financial performance.
- When the return on asset is used, the p value remains below 0.1 and the adjusted r square is relatively higher, with a lower residual error
- Regression with Return on Assets ratio: P-value < 0.1 shows a statistically significant relationship, and 38.8% of the variability in the DV is explained.

## Model 3 with Fixed effects

**Return on Asset= 0.000001065 \* RSTKGRNT + 0.000042918 \* total asset  
- 0.000997851\* profit margin + 0.000872509\* asset turnover**

# OPTION\_AWARDS\_BLK\_VALUE vs Firm Financial Performance

IV: OPTION_AWARDS_BLK_VALUE, DV: Net profit margin		IV: OPTION_AWARDS_BLK_VALUE, DV: Return on asset	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0147	Adj. R squared = 0.329	Adj. R squared = 0.0142	Adj. R squared = 0.39
RSE = 0.1409	RSE = 0.116	RSE = 0.0893	RSE = 0.0703
OABV(slope) = -0.000000499	OABV (slope) = -0.0000005164	OABV (slope) = -0.0000002347	OABV (slope) = -0.0000002893
OABV (P value) < 0.1	OABV (P value) > 0.1	OABV (P value) > 0.1	OABV (P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 12277		AIC = 21985	
RMSE = 0.139		RMSE = 0.08772	

- Model 1 and 2 have a very low adjusted R Squared value compared to Model 3
- The P-Value for Model 3 shows it is not statistically significant, unless we look at the F Statistic
- Slope is very small and negative, indicating weak association.

- Similarly, Model 1 and 2 have a lower Adj R. Squared Value compared to Model 3, and Model 3 is the only one with a P-Value less than 0.1
- Slope is very small and negative, indicating weak association
- R Square and RSE improve compared to profit margin.

# OPTION\_AWARDS\_BLK\_VALUE vs Firm Financial Performance

IV: OPTION_AWARDS_BLK_VALUE, DV: Current Ratio	
Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0924	Adj. R squared = 0.717
RSE = 1.43	RSE = 0.796
OABV (slope) = -0.0000004	OABV (slope) = 0.000003444
OABV (P value) > 0.1	OABV(P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 61428	
RMSE = 1.449	

- Improvement in r square when fixed effects are added
- P-Value remains statistically significant
- Slope changes from negative to positive when considering fixed effects

IV: OPTION_AWARDS_BLK_VALUE, DV: Debt to Equity	
Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.4618	Adj. R squared = 0.795
RSE = 1.886	RSE = 1.17
OABV (slope) = 0.000001245	OABV (slope) = -0.00000256
OABV (P value) > 0.1	OABV(P value) < 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 71145	
RMSE = 1.996	

- Highest Adjusted R Square out of the four regressions at 0.795.
- Highest RMSE and AIC, indicating weaker regression fit.
- Best model to determine firm financial performance in terms of R Squared and P-Value.

# OPTION\_AWARDS\_BLK\_VALUE vs Firm Financial Performance

## Summary

- Slope coefficient close to 0 shows a weak association between OPTION\_AWARDS\_BLK\_VALUE and firm financial performance.
- Fixed effects have effect on the independent variable, however given the small slope coefficient, there is a weak negative relationship between the IV and Net Profit Margin, Return on Assets and Debt to Equity Ratio, and a weak positive relationship with Current ratio

## Model 3 with Fixed effects

$$\begin{aligned}\text{Debt-to-equity-ratio} = & -0.00000256 * \text{OPTION\_AWARDS\_BLK\_VALUE} \\ & + 0.00000391 * \text{total asset} -0.62493929 * \text{net profit margin} + 0.76633488 * \text{asset turnover}\end{aligned}$$

# LTIP vs Firm Financial Performance

IV: LTIP/Total asset, DV: Net Profit Margin		IV: LTIP/Total asset, DV: Return on asset	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0145	Adj. R squared = 0.328	Adj. R squared = 0.0144	Adj. R squared = 0.39
RSE = 0.141	RSE = 0.116	RSE = 0.0893	RSE = 0.0702
LTIP_at (slope) = 0.0053578	LTIP_at (slope) = - 0.001826	LTIP_at (slope) = 0.0085003	LTIP_at(slope) = 0.0003515
LTIP_at (P value) > 0.1	LTIP_at (P value) > 0.1	LTIP_at (P value) < 0.1	LTIP_at (P value) > 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = -67884		AIC = -83701	
RMSE = 0.139		RMSE = 0.08725	

- Negative slope coefficient is statistically insignificant
- Firm and year fixed effect improved adj. r-squared, though remains low
- IV coefficient shifted to negative in Model 3

- Adjusted r squared remains relatively low across models
- P value for LTIP\_at indicates that LTIP is unlikely to be a significant predictor of firm's operation efficiency
- P value for full model implies that model as a whole is statistically significant

# LTIP vs Firm Financial Performance

IV: LTIP/Total asset, DV: Current Ratio		IV: LTIP/Total asset, DV: Debt to Equity Ratio	
Model 1 and 2	Model 3 (with Fixed Effects)	Model 1 and 2	Model 3 (with Fixed Effects)
Adj. R squared = 0.0932	Adj. R squared = 0.717	Adj. R squared = 0.462	Adj. R squared = 0.795
RSE = 1.42	RSE = 0.796	RSE = 1.89	RSE = 1.17
LTIP_at (slope) = - 0.230164	LTIP_at(slope) = - 0.068259	LTIP_at (slope) = 0.033826	LTIP_at (slope) = 0.006029
LTIP_at (P value) < 0.1	LTIP_at (P value) > 0.1	LTIP_at (P value) > 0.1	LTIP_at (P value) > 0.1
Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1	Full model (P value) < 0.1
AIC = 12262		AIC = 21983	
RMSE = 1.448		RMSE = 1.995	

- Outperforms previous models with net profit margin and return on assets
- Statistically significant negative relationship between long term incentive pay and current ratio

- Slope coefficient indicates a positive relationship
- With fixed effects, explanatory power of model is improved but p-value above threshold, potential lack of statistical significance

# LTIP vs Firm Financial Performance

## Summary

- Weak relationship indicated by a low adjusted r-squared and negative coefficient close to 0
- F statistic associated with the overall model is less than 0.1, indicating that the model is statistically significant.
- With P value for LITP\_ at more than 0.1 for all models, **LITP is unlikely to be a significant predictor of firm's financial performance**

# Summary

	Profitability (Net Profit Margin)	Operation Efficiency (Return on Assets)	Liquidity (Current Ratio)	Leverage (Debt to Equity Ratio)
Salary				✓
Bonus			✓	
Other Compensation				✓
Restricted Stock Grant		✓		
Options Awarded				✓
Long Term Incentive Pay				

# Limitations and Conclusion

<b>Limitations</b>	<ul style="list-style-type: none"><li>• Missing Values (NA)</li><li>• Outliers</li><li>• Data size and complications</li></ul>
<b>Conclusion</b>	<ul style="list-style-type: none"><li>• Weak Relationship indicated by <b>slope close to 0</b></li><li>• Except for Risk Grant, all other model indicates a <b>negative relationship</b></li><li>• Therefore, we <b>reject our null hypothesis</b> that there is a positive relationship</li></ul>
<b>Future Uses</b>	<ul style="list-style-type: none"><li>• Future research can be conducted to identify effect of <b>financial performance on executive compensation</b></li><li>• Further research may consider <b>different fixed effects and control variables</b></li></ul>

# Thank You

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