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Environment reset (run first)

```
rm(list = ls())           # clear objects
graphics.off()           # close plots
cat("\014")               # clear console
```

Step 1: Computing the requested arithmetic and logical expressions

```
123 * 453
```

```
## [1] 55719
```

```
5^2 * 40
```

```
## [1] 1000
```

```
TRUE & FALSE
```

```
## [1] FALSE
```

```
TRUE | FALSE
```

```
## [1] TRUE
```

```
75 %% 10
```

```
## [1] 5
```

```
75 / 10
```

```
## [1] 7.5
```

Step 2: Creating a vector with c() function and storing it in variable “first_vector”

```
first_vector <- c(17, 12, -33, 5)  
first_vector
```

```
## [1] 17 12 -33 5
```

Step 3: Creating a 5-multiples vector with c() function and storing it in variable “counting_by_fives”

```
counting_by_fives <- c(5, 10, 15, 20, 25, 30, 35)  
counting_by_fives
```

```
## [1] 5 10 15 20 25 30 35
```

Step 4: Creating a descending vector from 20 to 1 using range operator “:” and storing it in “second_vector”

```
second_vector <- 20:1
second_vector
```

```
## [1] 20 19 18 17 16 15 14 13 12 11 10 9 8 7 6 5 4 3 2 1
```

Step 5: Creating a vector from 5 to 15 using range operator “:” and storing it in “counting_vector”

```
counting_vector <- 5:15
counting_vector
```

```
## [1] 5 6 7 8 9 10 11 12 13 14 15
```

Step 6: Creating a vector of grades and storing it in “grades”

```
grades <- c(96, 100, 85, 92, 81, 72)
grades
```

```
## [1] 96 100 85 92 81 72
```

Step 7: Adding 3 bonus points to grades and storing it in “bonus_points_added”

```
bonus_points_added <- grades + 3
bonus_points_added
```

```
## [1] 99 103 88 95 84 75
```

Step 8: Creating a vector of numbers 1–100 using range operator “:” and storing it in “one_to_one_hundred”

```
one_to_one_hundred <- 1:100
one_to_one_hundred
```

```
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
## [40] 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63
## [79] 79 80 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
```

Step 9: Performing addition, multiplication, and logical comparisons on “second_vector”

```
#'Reading "second_vector"
second_vector

## [1] 20 19 18 17 16 15 14 13 12 11 10 9 8 7 6 5 4 3 2 1

#'Adding 20 to "second_vector" elements
second_vector + 20

## [1] 40 39 38 37 36 35 34 33 32 31 30 29 28 27 26 25 24 23 22 21

#'Reading "second_vector"
second_vector

## [1] 20 19 18 17 16 15 14 13 12 11 10 9 8 7 6 5 4 3 2 1

#'Multiplying 20 to "second_vector" elements
second_vector * 20

## [1] 400 380 360 340 320 300 280 260 240 220 200 180 160 140 120 100 80 60 40 20

#'Reading "second_vector"
second_vector

## [1] 20 19 18 17 16 15 14 13 12 11 10 9 8 7 6 5 4 3 2 1

#'Checking if "second_vector" elements are greater or equal than 20
second_vector >= 20

## [1] TRUE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE FALSE

#'Reading "second_vector"
second_vector

## [1] 20 19 18 17 16 15 14 13 12 11 10 9 8 7 6 5 4 3 2 1

#'Checking if "second_vector" elements are not equal to 20
second_vector != 20

## [1] FALSE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE

#'Reading "second_vector"
second_vector

## [1] 20 19 18 17 16 15 14 13 12 11 10 9 8 7 6 5 4 3 2 1
```

Since we are only computing arithmetic and logical expressions and not updating the “second_vector”, its value remains unchanged

Step 10: Calculating sum of numbers from 1–100 and storing it in “total”

```
total <- sum(one_to_one_hundred)
total
```

```
## [1] 5050
```

Step 11: Calculating average of numbers from 1–100 and storing it in “average_value”

```
average_value <- mean(one_to_one_hundred)
average_value
```

```
## [1] 50.5
```

Step 12: Calculating median of numbers from 1–100 and storing it in “median_value”

```
median_value <- median(one_to_one_hundred)
median_value
```

```
## [1] 50.5
```

Step 13: Finding maximum value from 1–100 and storing it in “max_value”

```
max_value <- max(one_to_one_hundred)
max_value
```

```
## [1] 100
```

Step 14: Finding minimum value from 1–100 and storing it in “min_value”

```
min_value <- min(one_to_one_hundred)
min_value
```

```
## [1] 1
```

Step 15: Extracting first element from “second_vector” and storing it in “first_value”

```
first_value <- second_vector[1]
first_value
```

```
## [1] 20
```

Step 16: Extracting first three elements from “second_vector” and storing them in “first_three_values”

```
first_three_values <- second_vector[1:3]
first_three_values
```

```
## [1] 20 19 18
```

Step 17: Extracting 1st, 5th, 10th, and 11th elements from “second_vector” and storing in “vector_from_brackets”

```
vector_from_brackets <- second_vector[c(1, 5, 10, 11)]
vector_from_brackets
```

```
## [1] 20 16 11 10
```

Step 18: Using Boolean indexing to extract selected elements from “first_vector” and storing in “vector_from_boolean_brackets”

```
vector_from_boolean_brackets <- first_vector[c(FALSE, TRUE, FALSE, TRUE)]
vector_from_boolean_brackets
```

```
## [1] 12 5
```

Only elements where the index is True is filtered and stored in new variable “vector_from_boolean_brackets”

Step 19: Checking which elements in “second_vector” are greater than or equal to 10

```
second_vector >= 10
```

```
## [1] TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE TRUE FALSE FALSE FALSE FALSE FALSE
```

```
second_vector
```

```
## [1] 20 19 18 17 16 15 14 13 12 11 10 9 8 7 6 5 4 3 2 1
```

First 10 elements of the “second_vector” are greater or equal to 10, meanwhile the “second_vector” is unchanged

Step 20: Filtering “one_to_one_hundred” to keep only values ≥ 20

```
one_to_one_hundred[one_to_one_hundred >= 20]
```

```
## [1] 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40 41 42 43
## [41] 60 61 62 63 64 65 66 67 68 69 70 71 72 73 74 75 76 77 78 79 80 81 82 83
## [81] 100
```

```
one_to_one_hundred
```

```
## [1] 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
## [40] 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54 55 56 57 58 59 60 61 62 63
## [79] 79 80 81 82 83 84 85 86 87 88 89 90 91 92 93 94 95 96 97 98 99 100
```

Only 81 elements are selected only for viewing (and not updated) out of 100, with filter condition of “one_to_one_hundred” elements greater or equal to 20.

Step 21: Filtering “grades” to keep only values greater than 85 and storing in “lowest_grades_removed”

```
lowest_grades_removed <- grades[grades > 85]
lowest_grades_removed
```

```
## [1] 96 100 92
```

Step 22: Removing 3rd and 4th elements from “grades” and storing in “middle_grades_removed”

```
middle_grades_removed <- grades[-c(3, 4)]  
middle_grades_removed
```

```
## [1] 96 100 81 72
```

Step 23: Removing 5th and 10th elements from “second_vector” and storing in “fifth_vector”

```
fifth_vector <- second_vector[-c(5, 10)]  
fifth_vector
```

```
## [1] 20 19 18 17 15 14 13 12 10 9 8 7 6 5 4 3 2 1
```

Step 24: Setting seed and generating 10 random numbers between 0 and 1000 using runif(), storing in “random_vector”

```
set.seed(5)  
random_vector <- runif(n = 10, min = 0, max = 1000)  
random_vector
```

```
## [1] 200.2145 685.2186 916.8758 284.3995 104.6501 701.0575 527.9600 807.9352 956.5001 110.4530
```

Step 25: Calculating total of “random_vector” and storing in “sum_vector”

```
sum_vector <- sum(random_vector)  
sum_vector
```

```
## [1] 5295.264
```

Step 26: Calculating cumulative sum of “random_vector” and storing in “cumsum_vector”


```
cumsum_vector <- cumsum(random_vector)
cumsum_vector
```

```
## [1] 200.2145 885.4330 1802.3088 2086.7083 2191.3584 2892.4159 3420.3759 4228.3111 5184.8112 5295.1
```

Step 27: Calculating mean of “random_vector” and storing in “mean_vector”

```
mean_vector <- mean(random_vector)
mean_vector
```

```
## [1] 529.5264
```

Step 28: Calculating standard deviation of “random_vector” and storing in “sd_vector”

```
sd_vector <- sd(random_vector)
sd_vector
```

```
## [1] 331.3606
```

Step 29: Rounding values of “random_vector” to nearest integers and storing in “round_vector”

```
round_vector <- round(random_vector)
round_vector
```

```
## [1] 200 685 917 284 105 701 528 808 957 110
```

Step 30: Sorting values of “random_vector” in ascending order and storing in “sort_vector”

```
sort_vector <- sort(random_vector)
sort_vector
```

```
## [1] 104.6501 110.4530 200.2145 284.3995 527.9600 685.2186 701.0575 807.9352 916.8758 956.5001
```

Step 31: Downloading the datafile ds_salaries.csv from Canvas and Saving it in same directory as this project

Step 32: Reading dataset “ds_salaries.csv” into R and storing it as “first_dataframe”

```
first_dataframe <- read.csv("ds_salaries.csv")
first_dataframe
```

##	X	work_year	experience_level	employment_type	job_title	salary
## 1	0	2020	MI	FT	Data Scientist	70000
## 2	1	2020	SE	FT	Machine Learning Scientist	260000
## 3	2	2020	SE	FT	Big Data Engineer	85000
## 4	3	2020	MI	FT	Product Data Analyst	20000
## 5	4	2020	SE	FT	Machine Learning Engineer	150000
## 6	5	2020	EN	FT	Data Analyst	72000
## 7	6	2020	SE	FT	Lead Data Scientist	190000
## 8	7	2020	MI	FT	Data Scientist	1100000
## 9	8	2020	MI	FT	Business Data Analyst	135000
## 10	9	2020	SE	FT	Lead Data Engineer	125000
## 11	10	2020	EN	FT	Data Scientist	45000
## 12	11	2020	MI	FT	Data Scientist	3000000
## 13	12	2020	EN	FT	Data Scientist	35000
## 14	13	2020	MI	FT	Lead Data Analyst	87000
## 15	14	2020	MI	FT	Data Analyst	85000
## 16	15	2020	MI	FT	Data Analyst	8000
## 17	16	2020	EN	FT	Data Engineer	4450000
## 18	17	2020	SE	FT	Big Data Engineer	100000
## 19	18	2020	EN	FT	Data Science Consultant	423000
## 20	19	2020	MI	FT	Lead Data Engineer	56000
## 21	20	2020	MI	FT	Machine Learning Engineer	299000
## 22	21	2020	MI	FT	Product Data Analyst	450000
## 23	22	2020	SE	FT	Data Engineer	42000
## 24	23	2020	MI	FT	BI Data Analyst	98000
## 25	24	2020	MI	FT	Lead Data Scientist	115000
## 26	25	2020	EX	FT	Director of Data Science	325000
## 27	26	2020	EN	FT	Research Scientist	42000
## 28	27	2020	SE	FT	Data Engineer	720000
## 29	28	2020	EN	CT	Business Data Analyst	100000
## 30	29	2020	SE	FT	Machine Learning Manager	157000
## 31	30	2020	MI	FT	Data Engineering Manager	51999
## 32	31	2020	EN	FT	Big Data Engineer	70000
## 33	32	2020	SE	FT	Data Scientist	60000
## 34	33	2020	MI	FT	Research Scientist	450000
## 35	34	2020	MI	FT	Data Analyst	41000
## 36	35	2020	MI	FT	Data Engineer	65000
## 37	36	2020	MI	FT	Data Science Consultant	103000
## 38	37	2020	EN	FT	Machine Learning Engineer	250000
## 39	38	2020	EN	FT	Data Analyst	10000
## 40	39	2020	EN	FT	Machine Learning Engineer	138000
## 41	40	2020	MI	FT	Data Scientist	45760

##	42	41	2020	EX	FT	Data Engineering Manager	70000
##	43	42	2020	MI	FT	Machine Learning Infrastructure Engineer	44000
##	44	43	2020	MI	FT	Data Engineer	106000
##	45	44	2020	MI	FT	Data Engineer	88000
##	46	45	2020	EN	PT	ML Engineer	14000
##	47	46	2020	MI	FT	Data Scientist	60000
##	48	47	2020	SE	FT	Data Engineer	188000
##	49	48	2020	MI	FT	Data Scientist	105000
##	50	49	2020	MI	FT	Data Engineer	61500
##	51	50	2020	EN	FT	Data Analyst	450000
##	52	51	2020	EN	FT	Data Analyst	91000
##	53	52	2020	EN	FT	AI Scientist	300000
##	54	53	2020	EN	FT	Data Engineer	48000
##	55	54	2020	SE	FL	Computer Vision Engineer	60000
##	56	55	2020	SE	FT	Principal Data Scientist	130000
##	57	56	2020	MI	FT	Data Scientist	34000
##	58	57	2020	MI	FT	Data Scientist	118000
##	59	58	2020	SE	FT	Data Scientist	120000
##	60	59	2020	MI	FT	Data Scientist	138350
##	61	60	2020	MI	FT	Data Engineer	110000
##	62	61	2020	MI	FT	Data Engineer	130800
##	63	62	2020	EN	PT	Data Scientist	19000
##	64	63	2020	SE	FT	Data Scientist	412000
##	65	64	2020	SE	FT	Machine Learning Engineer	40000
##	66	65	2020	EN	FT	Data Scientist	55000
##	67	66	2020	EN	FT	Data Scientist	43200
##	68	67	2020	SE	FT	Data Science Manager	190200
##	69	68	2020	EN	FT	Data Scientist	105000
##	70	69	2020	SE	FT	Data Scientist	80000
##	71	70	2020	MI	FT	Data Scientist	55000
##	72	71	2020	MI	FT	Data Scientist	37000
##	73	72	2021	EN	FT	Research Scientist	60000
##	74	73	2021	EX	FT	BI Data Analyst	150000
##	75	74	2021	EX	FT	Head of Data	235000
##	76	75	2021	SE	FT	Data Scientist	45000
##	77	76	2021	MI	FT	BI Data Analyst	100000
##	78	77	2021	MI	PT	3D Computer Vision Researcher	400000
##	79	78	2021	MI	CT	ML Engineer	270000
##	80	79	2021	EN	FT	Data Analyst	80000
##	81	80	2021	SE	FT	Data Analytics Engineer	67000
##	82	81	2021	MI	FT	Data Engineer	140000
##	83	82	2021	MI	FT	Applied Data Scientist	68000
##	company_location		company_size				
##	1		DE		L		
##	2		JP		S		
##	3		GB		M		
##	4		HN		S		
##	5		US		L		
##	6		US		L		
##	7		US		S		
##	8		HU		L		
##	9		US		L		
##	10		NZ		S		
##	11		FR		S		

## 12	IN	L
## 13	FR	M
## 14	US	L
## 15	US	L
## 16	PK	L
## 17	JP	S
## 18	GB	S
## 19	IN	M
## 20	US	M
## 21	CN	M
## 22	IN	L
## 23	GR	L
## 24	US	M
## 25	AE	L
## 26	US	L
## 27	NL	L
## 28	MX	S
## 29	US	L
## 30	CA	L
## 31	DE	S
## 32	US	L
## 33	US	L
## 34	US	M
## 35	FR	L
## 36	AT	L
## 37	US	L
## 38	US	L
## 39	NG	S
## 40	US	S
## 41	US	S
## 42	ES	L
## 43	PT	M
## 44	US	L
## 45	GB	L
## 46	DE	S
## 47	GB	S
## 48	US	L
## 49	US	L
## 50	FR	L
## 51	IN	S
## 52	US	L
## 53	DK	S
## 54	DE	L
## 55	US	S
## 56	DE	M
## 57	ES	M
## 58	US	M
## 59	US	L
## 60	US	M
## 61	US	L
## 62	US	M
## 63	IT	S
## 64	US	L
## 65	HR	S

```
## 66          DE          S
## 67          DE          S
## 68          US          M
## 69          US          S
## 70          AT          S
## 71          LU          S
## 72          FR          S
## 73          GB          L
## 74          US          L
## 75          US          L
## 76          FR          L
## 77          US          M
## 78          IN          M
## 79          US          L
## 80          US          M
## 81          DE          L
## 82          US          L
## 83          CA          L
## [ reached 'max' / getOption("max.print") -- omitted 524 rows ]
```

Step 33: Generating summary statistics of the dataset “first_dataframe”

```
summary(first_dataframe)
```

```
##          X          work_year  experience_level  employment_type  job_title          salary
## Min.   : 0.0   Min.   :2020   Length:607      Length:607      Length:607      Min.   :
## 1st Qu.:151.5  1st Qu.:2021   Class :character  Class :character  Class :character  1st Qu.:
## Median :303.0  Median :2022   Mode  :character  Mode  :character  Mode  :character  Median : 1
## Mean   :303.0  Mean   :2021                                     Mean   : 3
## 3rd Qu.:454.5  3rd Qu.:2022                                     3rd Qu.: 1
## Max.   :606.0  Max.   :2022                                     Max.   :304
## remote_ratio  company_location  company_size
## Min.   : 0.00   Length:607      Length:607
## 1st Qu.: 50.00   Class :character  Class :character
## Median :100.00   Mode  :character  Mode  :character
## Mean   : 70.92
## 3rd Qu.:100.00
## Max.   :100.00
```

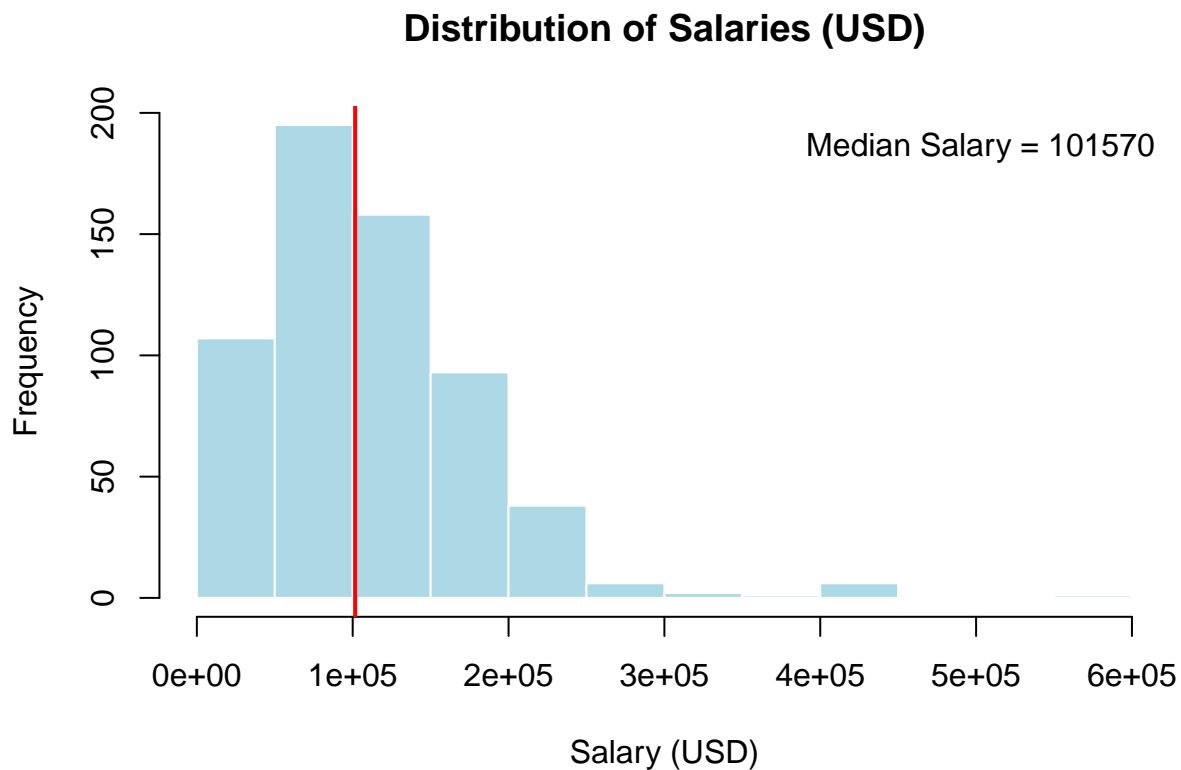
Visualizations + Recommendations

1. Salary Distribution (Histogram)

Reason: Helps visualize how salaries are spread, showing skewness and outliers.

```
hist(first_dataframe$salary_in_usd,
     main = "Distribution of Salaries (USD)",
     xlab = "Salary (USD)",
```

```
col = "lightblue",
border = "white")
abline(v = median(first_dataframe$salary_in_usd, na.rm = TRUE),
      col = "red", lwd = 2)
legend("topright", legend = c(paste("Median Salary =", round(median(first_dataframe$salary_in_usd, na.rm = TRUE), 0)),
                             bty = "n")
```



2. Salary by Experience Level (Boxplot)

Reason: Shows how pay changes with experience, revealing career-level pay gaps.

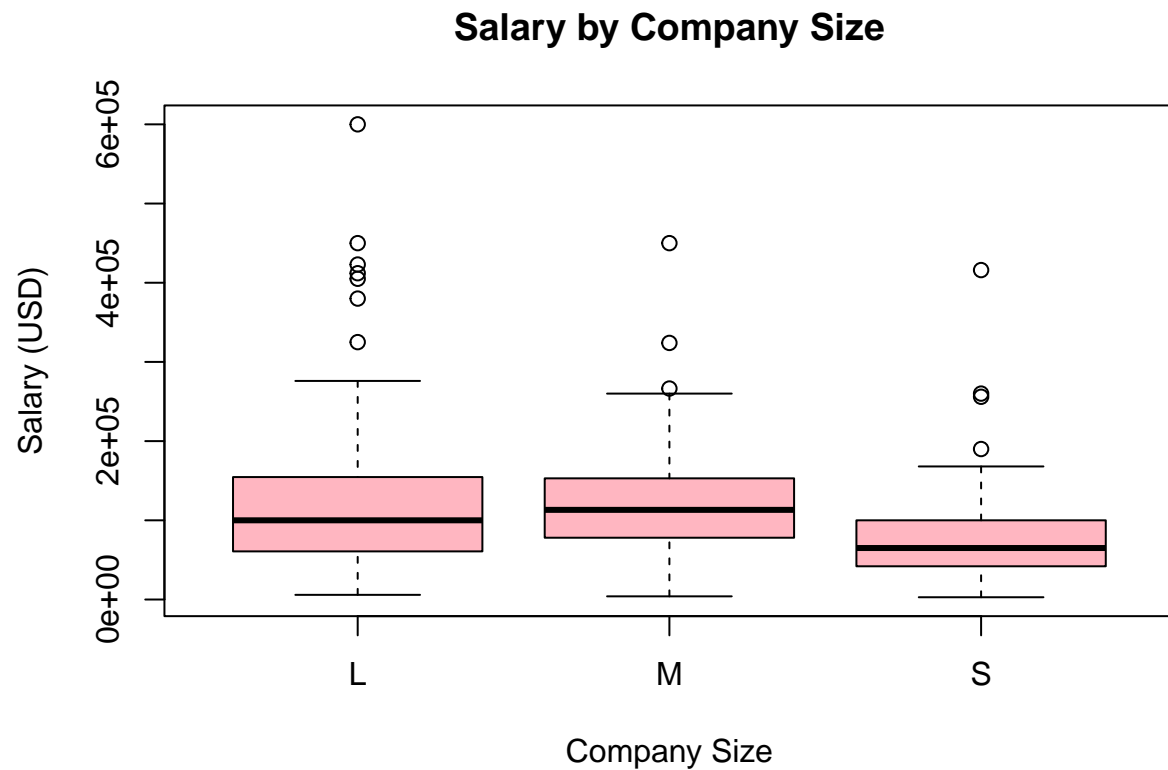
```
boxplot(salary_in_usd ~ experience_level, data = first_dataframe,
        main = "Salary by Experience Level",
        xlab = "Experience Level",
        ylab = "Salary (USD)",
        col = "lightgreen")
```



3. Salary by Company Size (Boxplot)

Reason: Highlights differences in pay trends across small, medium, and large companies.

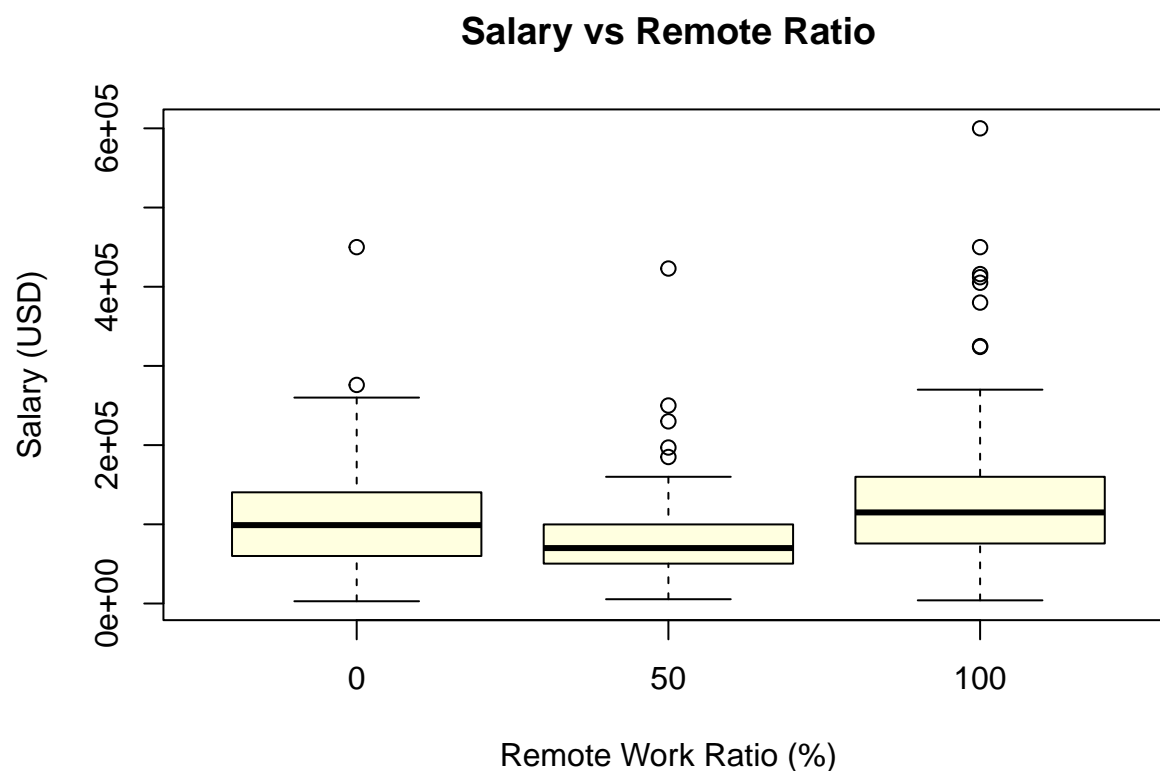
```
boxplot(salary_in_usd ~ company_size, data = first_dataframe,  
        main = "Salary by Company Size",  
        xlab = "Company Size",  
        ylab = "Salary (USD)",  
        col = "lightpink")
```



4. Remote Ratio vs Salary (Boxplot)

Reason: Shows how compensation differs between on-site, hybrid, and fully remote roles.

```
boxplot(salary_in_usd ~ remote_ratio, data = first_dataframe,  
        main = "Salary vs Remote Ratio",  
        xlab = "Remote Work Ratio (%)",  
        ylab = "Salary (USD)",  
        col = "lightyellow")
```

Findings & Insights

The visualizations show that salaries in the dataset are right-skewed, with most professionals earning around the median while a few high-paying roles drive up the average. Pay clearly increases with experience, moving from entry- to executive-level positions. Larger companies tend to offer higher compensation than small or mid-sized ones, reflecting resource differences. Finally, remote roles show comparable or slightly higher pay, suggesting flexibility is increasingly valued without reducing earnings potential.

References

R Core Team (2025). *R: A Language and Environment for Statistical Computing*.
 Xie, Y. (2023). *knitr: A General-Purpose Package for Dynamic Report Generation in R*.
 All code written by the author for Yadav_Project1.

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