

NBA SALARIES /STATISTICAL MODELS

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Two-sided test on the mean:

- Equal to \$5m
- Greater than \$4m

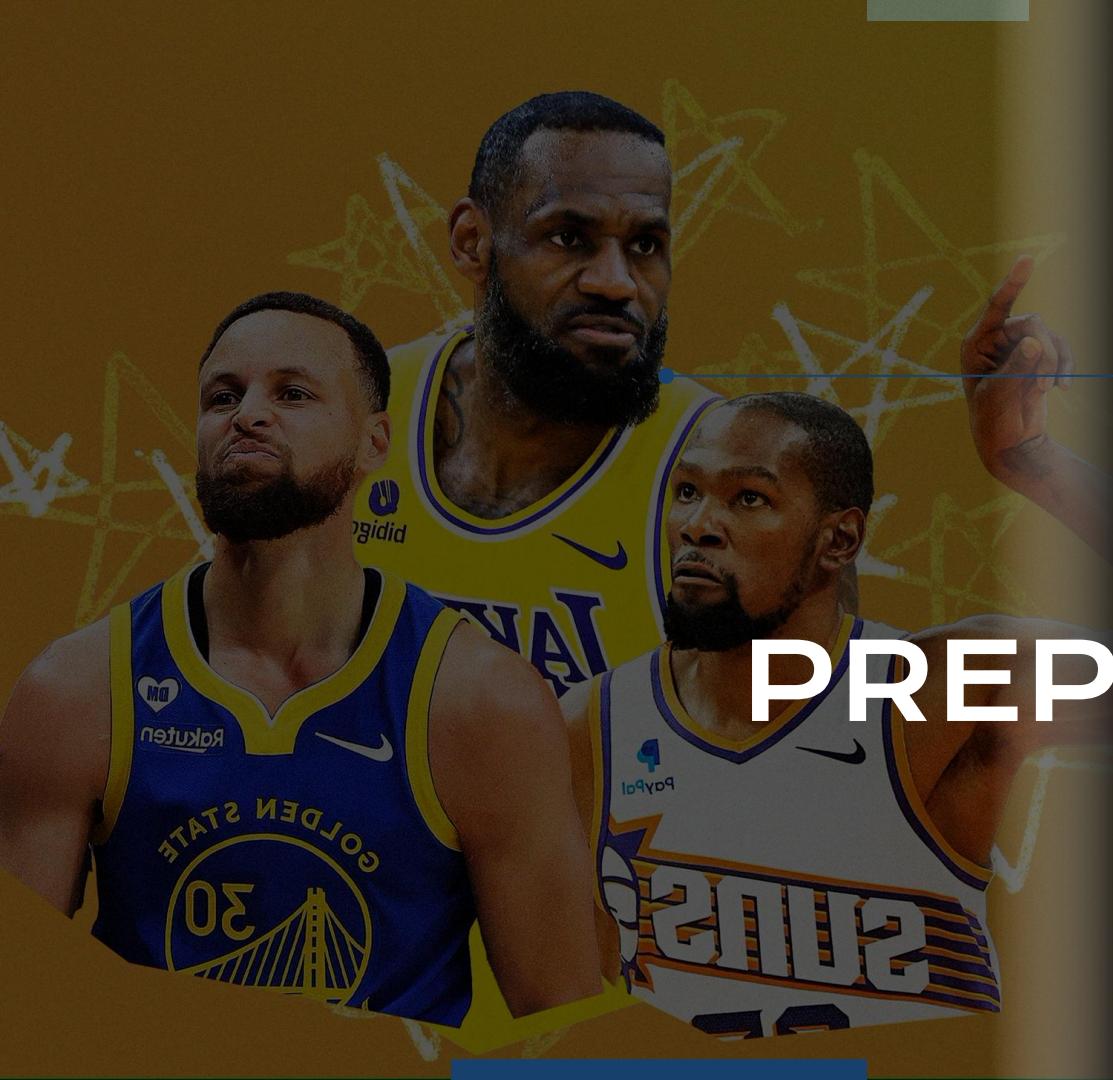
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Linear Regression

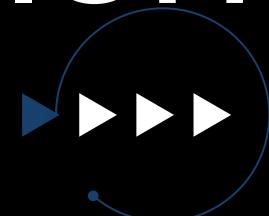
Two-sided test on the mean:

- Equal to \$5m
- Greater than \$4m



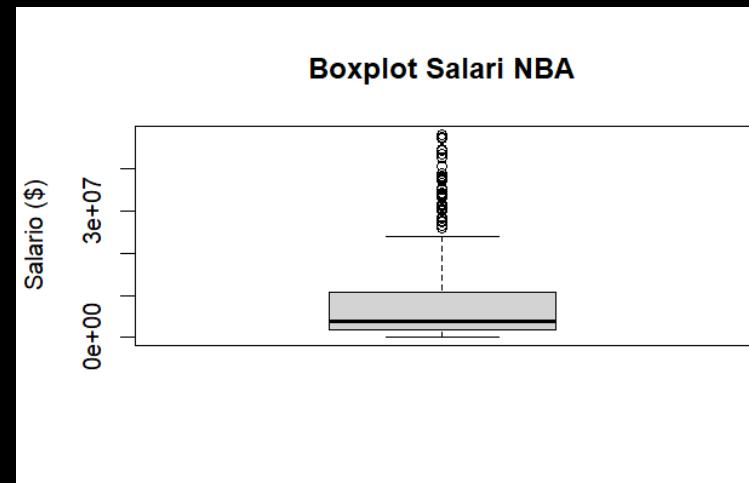
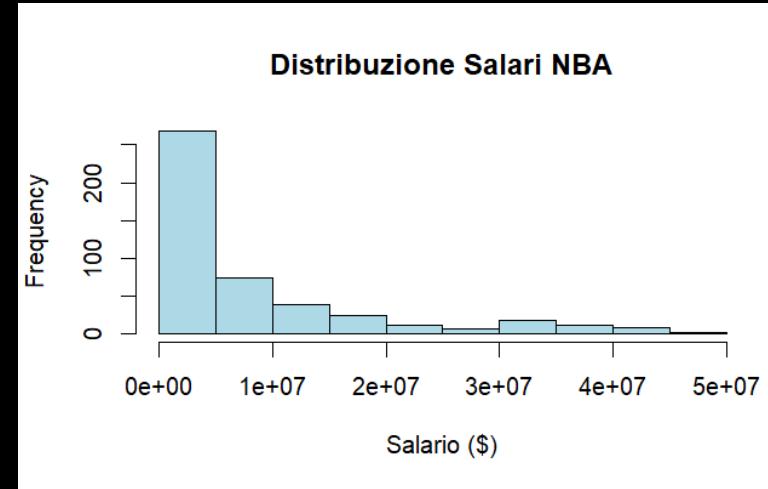


/01 DATA PREPARATION



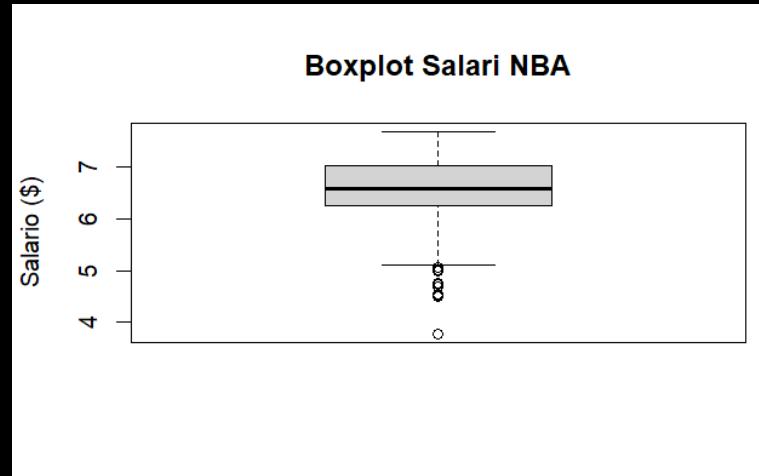
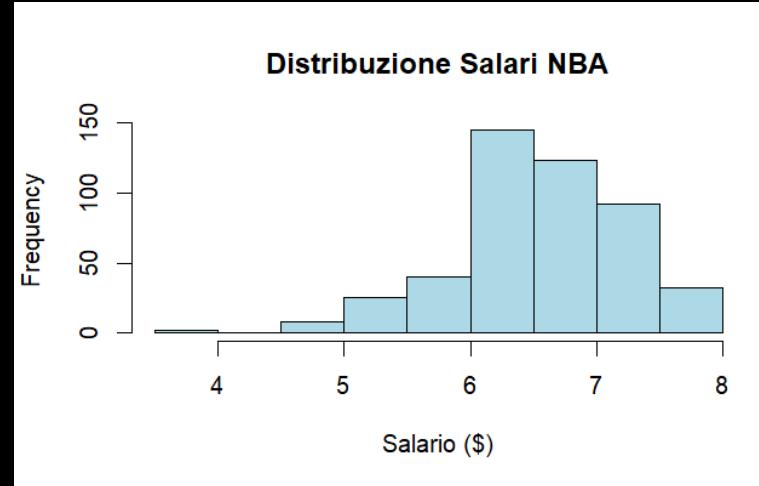
WITHOUT USING LOG10()

- NBA salaries have a very asymmetrical distribution (few earn a lot and many earn little).
- Less readable graphs.
- Mean influenced by outliers.



USING LOG10()

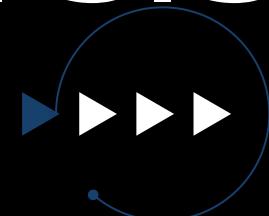
- More symmetric and linear distribution.
- Outliers are more contained.
- Data more suitable for building regression models.





/02

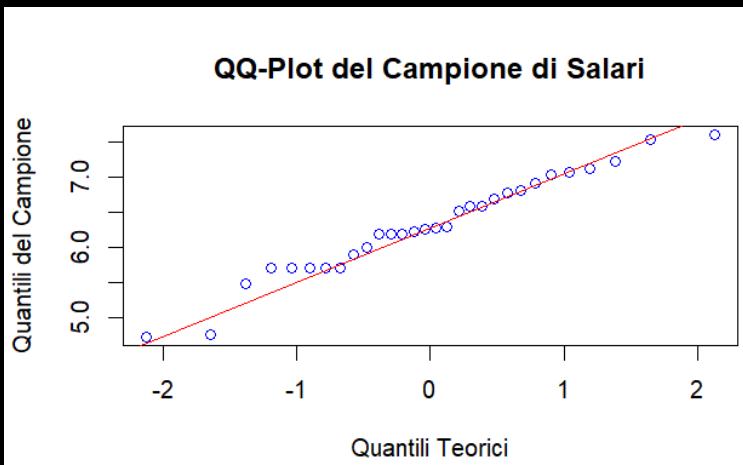
DESCRIPTIVE ANALYSIS



SHAPIRO-WILK TEST

- **Hypothesis H0 (null):** The sample follows a normal distribution.
- **H1 (alternative):** It does not follow a linear distribution.
- **P-value > 0.05:** Do not reject H0.
- **QQ-Plot:** The data follows the "perfect normality" line quite well but with slight deviations at the extremes. (Graph consistent with the Test).

```
Shapiro-Wilk normality test  
data: sample_salary  
W = 0.97083, p-value = 0.5623
```



CONFIDENCE INTERVAL

The mean salary is between 6.279 and 6.823 on a logarithmic scale.

With monetary conversion, we are talking about an interval between approximately \$2m and \$6.5m.

High salaries but with high variability.

```
mean_salary <- mean(sample_salary) |  
sd_salary <- sd(sample_salary)  
n <- length(sample_salary)  
error_std <- qt(0.975, df = n-1) * sd_salary / sqrt(n)  
  
conf_interval <- c(mean_salary - error_log, mean_salary + error_log)  
conf_interval
```

A composite image featuring two NBA players. On the left, LeBron James in a yellow Los Angeles Lakers jersey (number 23) looks intensely at the camera. On the right, Stephen Curry in a Golden State Warriors jersey (number 30) is captured mid-shout with his mouth wide open. The background is a dark, blurred basketball court.

/03

HYPOTHESIS TESTING

HIGHEST PAID PLAYERS 2024

PRESENTED BY SPORTICO



Is the Mean Salary different from \$5m? /t-test

One Sample t-test

```
data: sample_salary
t = -2.9426, df = 29, p-value = 0.006342
alternative hypothesis: true mean is not equal to 6.69897
95 percent confidence interval:
 6.052179 6.582601
sample estimates:
mean of x
 6.31739
```

H0: The mean **is equal** to 5,000,000.

H1: The mean is different from 5,000,000.

P-value = 0.006: Very small -> reject H0 (the sample mean of salaries is significantly different from \$5m)

The confidence interval does not include 6.69897, confirming the result.

Is the Mean Salary greater than \$4m? /t-test

One Sample t-test

```
data: sample_salary
t = -2.1953, df = 29, p-value = 0.9819
alternative hypothesis: true mean is greater than 6.60206
95 percent confidence interval:
 6.097059      Inf
sample estimates:
mean of x
 6.31739
```

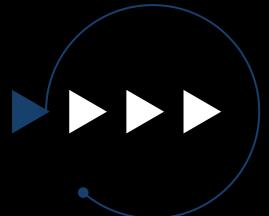
H0: The mean **is less** than or equal to 4,000,000.

H1: The mean **is greater** than 4,000,000.

P-value = 0.9819: Much larger than 0.05 -> do not reject H0 (we do not have sufficient evidence to say that the mean salary is greater than \$4m).



/03 LINEAR REGRESSION TEST



DOES TS% INFLUENCE SALARY?

Very low P-value: High significance.

Low R^2: Salary is influenced by TS% but only to a small extent.

If TS% increased by 10%?

$$1.0947 * 0.01 = 0.10947 \rightarrow 10^{0.10947} \rightarrow 1.29$$

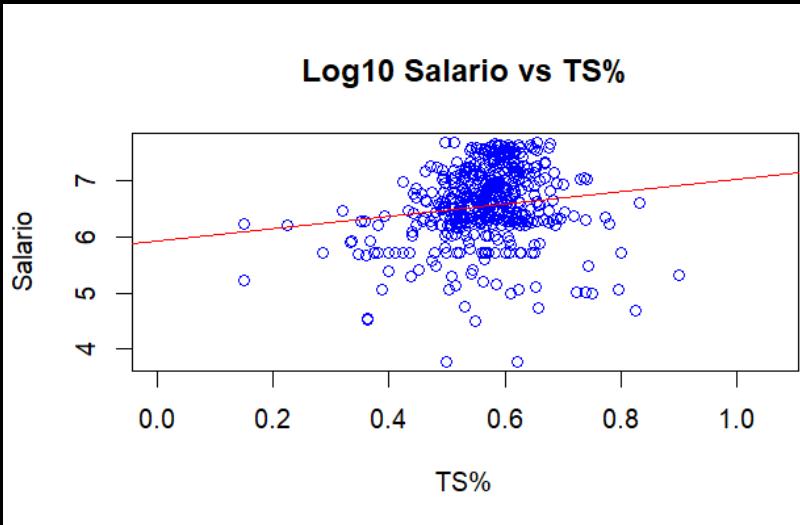
The salary would therefore increase by 29%.

Assuming a player earns 5 million, with a 10% increase in TS%, the player will earn 6.45 million dollars (+29%).

Coefficients:

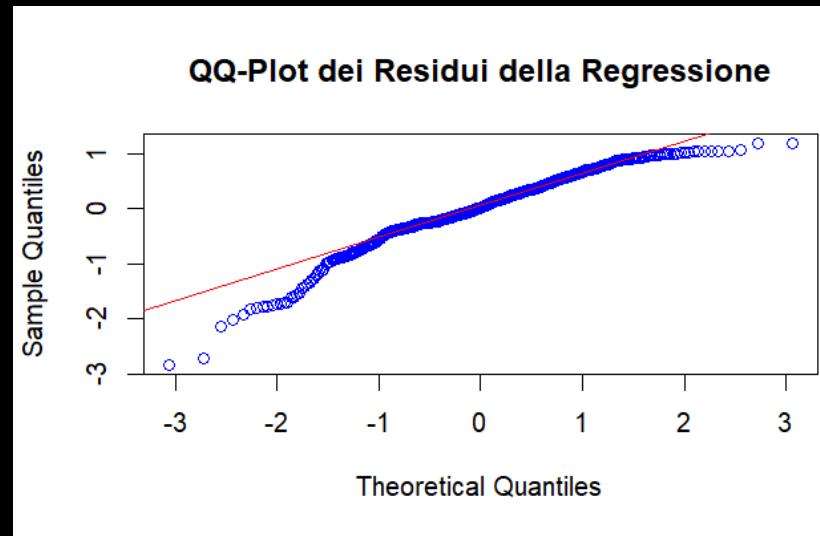
	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	5.9320	0.1742	34.062	< 2e-16 ***	
`TS%`	1.0947	0.3045	3.595	0.000359 ***	

Signif. codes:	0 '***'	0.001 '**'	0.01 '*'	0.05 '.'	0.1 ' '
Residual standard error:	0.6615	on 464 degrees of freedom			
(1 osservazione eliminata a causa di un valore mancante)					
Multiple R-squared:	0.0271,	Adjusted R-squared:	0.02501		
F-statistic:	12.93	on 1 and 464 DF,	p-value:	0.0003588	



REGRESSION RESIDUALS: Salary vs TS%

- The residuals are distributed along the theoretical line in the central part, indicating a good approximation to normality.
- Some slight deviations at the extremes are present, but not such that they compromise the reliability of the model.



DOES AGE INFLUENCE SALARY?

Very low P-value: High significance.

Low R^2: Salary is influenced by age but only to a small extent.

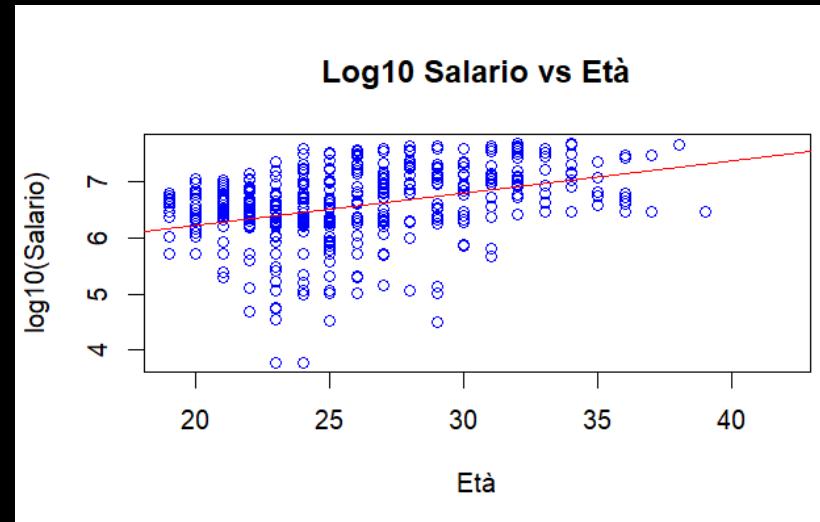
Every year of age brings a salary increase of 0.057.
 $10^{0.057} = 1.14 \rightarrow 14\%$

If a player earns \$5m, the following year (if still in the league), they will earn \$5.7m (+14%)

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	Signif. codes:
(Intercept)	5.066438	0.177423	28.556	< 2e-16 ***	0 ‘***’ 0.001 ‘**’ 0.01 ‘*’ 0.05 ‘.’ 0.1 ‘ ’ 1
Age	0.057300	0.006779	8.452	3.71e-16 ***	

Residual standard error: 0.6256 on 465 degrees of freedom
Multiple R-squared: 0.1332, Adjusted R-squared: 0.1313
F-statistic: 71.44 on 1 and 465 DF, p-value: 3.712e-16

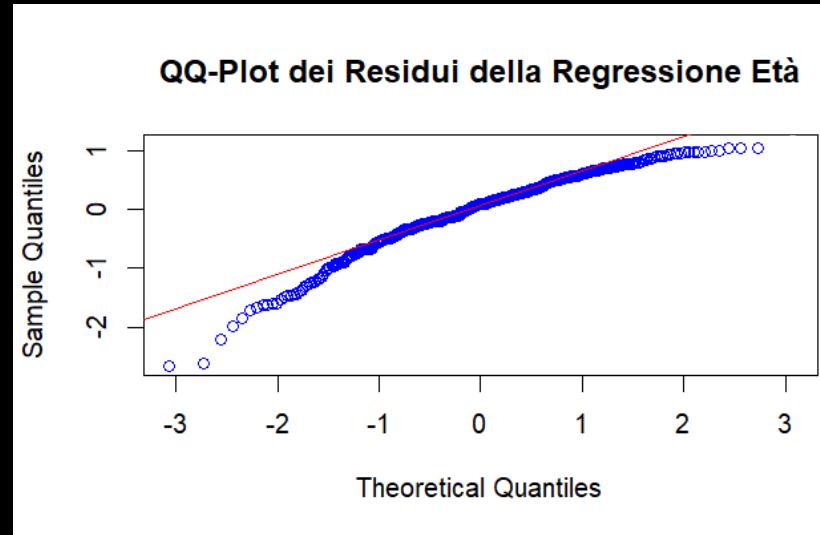


REGRESSION RESIDUALS: Salary vs Age

The QQ-Plot of residuals on the Age regression is very similar to the one seen for TS%.

The residuals follow the line quite well in the center.

At the extremes (especially for very high or very low values) there is some deviation.





THANKS FOR
YOUR
ATTENTION