# STAT632 Project: Advertisement Sales Modeling

```
# --- 1. Load Required Libraries ---
library(MASS)
                   # Box-Cox transformation
library(glmnet)
                      # LASSO regression
Loading required package: Matrix
Loaded glmnet 4.1-8
library(randomForest) # Random Forest
randomForest 4.7-1.2
Type rfNews() to see new features/changes/bug fixes.
library(car)
                       # Multicollinearity (VIF)
Loading required package: carData
library(ggplot2)
                       # Data visualization
Attaching package: 'ggplot2'
The following object is masked from 'package:randomForest':
    margin
```

```
library(caret)
                       # Train/test split and validation
Loading required package: lattice
library(dplyr)
                       # Data manipulation
Attaching package: 'dplyr'
The following object is masked from 'package:car':
    recode
The following object is masked from 'package:randomForest':
    combine
The following object is masked from 'package:MASS':
    select
The following objects are masked from 'package:stats':
    filter, lag
The following objects are masked from 'package:base':
    intersect, setdiff, setequal, union
# --- 2. Load and Explore Data ---
adver <- read.csv("Advertising And Sales.csv")</pre>
summary(adver)
```

ID		TV	Radio	Newspaper
Min.	: 1.00	Min. : 0.70	Min. : 0.00	Min. : 0.30
1st Qu	.: 50.75	1st Qu.: 74.38	1st Qu.:10.07	1st Qu.: 12.75
Median	:100.50	Median :149.75	Median :22.90	Median : 25.75
Mean	:100.50	Mean :147.03	Mean :23.29	Mean : 30.55

#### str(adver)

```
'data.frame': 200 obs. of 5 variables:

$ ID : int 1 2 3 4 5 6 7 8 9 10 ...

$ TV : num 230.1 44.5 17.2 151.5 180.8 ...

$ Radio : num 37.8 39.3 45.9 41.3 12.8 48.9 32.8 19.6 2.1 2.6 ...

$ Newspaper: num 69.2 45.1 69.3 58.5 58.4 75 23.5 11.6 1 21.2 ...

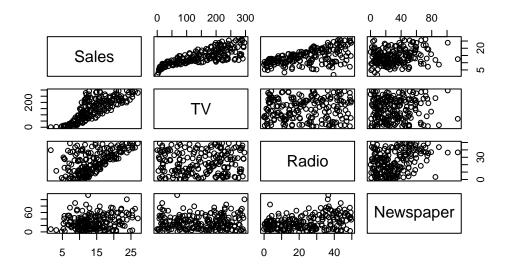
$ Sales : num 22.1 10.4 9.3 18.5 12.9 7.2 11.8 13.2 4.8 10.6 ...

# --- 3. Pairwise Scatterplot ---

pairs(Sales ~ TV + Radio + Newspaper, data = adver,

main = "Pairwise Scatterplot of Sales and Advertising Channels")
```

## Pairwise Scatterplot of Sales and Advertising Channels



```
# --- 4. Remove Outliers ---
adver \leftarrow adver[-c(131, 156, 99, 108, 200), ]
# --- 5. Check Missing Values ---
colSums(is.na(adver))
      ID
               TV
                      Radio Newspaper
                                        Sales
# --- 6. Base Linear Model (TV, Radio, Newspaper) ---
lm1 <- lm(Sales ~ TV + Radio + Newspaper, data = adver)</pre>
summary(lm1)
Call:
lm(formula = Sales ~ TV + Radio + Newspaper, data = adver)
Residuals:
   Min
            1Q Median
                          3Q
                                 Max
-5.4236 -0.9315 0.1920 1.1625 2.7446
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 3.171631 0.296987 10.679 <2e-16 ***
TV
            Radio
            Newspaper -0.005430 0.005479 -0.991 0.323
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.558 on 191 degrees of freedom
Multiple R-squared: 0.907, Adjusted R-squared: 0.9055
F-statistic: 620.8 on 3 and 191 DF, p-value: < 2.2e-16
# --- 7. Reduced Linear Model (TV and Radio only) ---
lm2 <- lm(Sales ~ TV + Radio, data = adver)</pre>
summary(lm2)
```

Call:

```
lm(formula = Sales ~ TV + Radio, data = adver)
Residuals:
   Min
            1Q Median
                            3Q
                                   Max
-5.5951 -0.8560 0.2263 1.1425 2.7664
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 3.073673 0.280043
                                 10.98
                                        <2e-16 ***
           0.044597
                                         <2e-16 ***
                      0.001321
                                 33.77
           0.190869
                      0.007583
                                 25.17
Radio
                                        <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 1.558 on 192 degrees of freedom
Multiple R-squared: 0.9065,
                               Adjusted R-squared: 0.9055
F-statistic: 930.8 on 2 and 192 DF, p-value: < 2.2e-16
# --- 8. Interaction Model (TV * Radio * Newspaper) ---
lm_interaction <- lm(Sales ~ TV * Radio * Newspaper, data = adver)</pre>
summary(lm_interaction)
Call:
lm(formula = Sales ~ TV * Radio * Newspaper, data = adver)
Residuals:
   Min
          10 Median
                        3Q
                              Max
-2.820 -0.371 0.144 0.488 1.458
Coefficients:
                    Estimate Std. Error t value Pr(>|t|)
                   6.818e+00 4.028e-01 16.926 < 2e-16 ***
(Intercept)
TV
                   1.844e-02 2.346e-03 7.859 2.96e-13 ***
                   3.435e-02 1.425e-02 2.410
Radio
                                                0.0169 *
Newspaper
                   1.293e-03 1.462e-02 0.088 0.9297
TV:Radio
                   1.089e-03 8.411e-05 12.950 < 2e-16 ***
                  -3.634e-06 7.919e-05 -0.046 0.9635
TV:Newspaper
Radio:Newspaper
                  -4.164e-05 4.108e-04 -0.101 0.9194
TV:Radio:Newspaper -3.037e-07 2.294e-06 -0.132 0.8948
```

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.05 '.' 0.1 ' ' 1

```
Residual standard error: 0.7803 on 187 degrees of freedom Multiple R-squared: 0.9772, Adjusted R-squared: 0.9763 F-statistic: 1143 on 7 and 187 DF, p-value: < 2.2e-16
```

#### Call:

```
 lm(formula = Sales ~ TV + I(TV^2) + Radio + I(Radio^2) + Newspaper + I(Newspaper^2) + I(TV^3) + I(TV^4) + I(TV^5) + TV * Radio * Newspaper, data = adver)
```

#### Residuals:

Min 1Q Median 3Q Max -0.84050 -0.20301 -0.01117 0.18327 0.84042

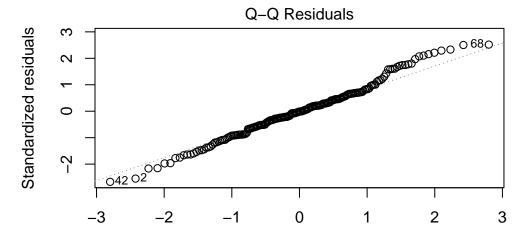
#### Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	3.594e+00	2.256e-01	15.930	< 2e-16	***
TV	1.500e-01	1.175e-02	12.768	< 2e-16	***
I(TV^2)	-1.677e-03	2.420e-04	-6.931	7.10e-11	***
Radio	3.178e-02	7.723e-03	4.114	5.89e-05	***
I(Radio^2)	1.954e-04	1.333e-04	1.466	0.144304	
Newspaper	-1.477e-04	6.455e-03	-0.023	0.981768	
<pre>I(Newspaper^2)</pre>	-5.305e-08	4.129e-05	-0.001	0.998976	
I(TV^3)	1.020e-05	2.033e-06	5.017	1.25e-06	***
I(TV^4)	-3.028e-08	7.478e-09	-4.049	7.61e-05	***
I(TV^5)	3.453e-11	9.984e-12	3.459	0.000677	***
TV:Radio	1.057e-03	3.816e-05	27.706	< 2e-16	***
TV:Newspaper	4.602e-06	3.615e-05	0.127	0.898857	
Radio:Newspaper	7.962e-05	2.004e-04	0.397	0.691651	
TV:Radio:Newspaper	-6.359e-07	1.081e-06	-0.588	0.557261	
Signif. codes: 0	'***' 0.001	'**' 0.01 <sup>'</sup>	'*' 0.05	'.' 0.1 '	' ' 1

Residual standard error: 0.3267 on 181 degrees of freedom Multiple R-squared: 0.9961, Adjusted R-squared: 0.9958

```
F-statistic: 3578 on 13 and 181 DF, p-value: < 2.2e-16
# --- 10. Reduced Polynomial + Interaction Model ---
lm_poly1 <- lm(Sales ~ TV + I(TV^2) + Radio + TV:Radio +</pre>
              I(TV^3) + I(TV^4) + I(TV^5), data = adver)
summary(lm poly1)
Call:
lm(formula = Sales ~ TV + I(TV^2) + Radio + TV:Radio + I(TV^3) +
    I(TV^4) + I(TV^5), data = adver)
Residuals:
     Min
                                3Q
                                        Max
              1Q
                   Median
-0.85876 -0.19498 -0.00409 0.17912 0.81061
Coefficients:
             Estimate Std. Error t value Pr(>|t|)
(Intercept) 3.541e+00 1.877e-01 18.865 < 2e-16 ***
            1.472e-01 1.145e-02 12.854 < 2e-16 ***
TV
I(TV^2)
           -1.610e-03 2.327e-04 -6.919 7.01e-11 ***
Radio
            4.484e-02 3.191e-03 14.051 < 2e-16 ***
I(TV^3)
            9.599e-06 1.952e-06 4.919 1.90e-06 ***
I(TV^4)
           -2.795e-08 7.164e-09 -3.901 0.000134 ***
            3.125e-11 9.536e-12 3.277 0.001250 **
I(TV^5)
TV:Radio
            1.030e-03 1.870e-05 55.094 < 2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 0.3255 on 187 degrees of freedom
Multiple R-squared: 0.996, Adjusted R-squared: 0.9959
F-statistic: 6698 on 7 and 187 DF, p-value: < 2.2e-16
```

# --- 11. Residual Diagnostics --plot(lm\_poly1, which = 2) # Q-Q plot



Theoretical Quantiles  $Im(Sales \sim TV + I(TV^2) + Radio + TV:Radio + I(TV^3) + I(TV^4) + I(TV^4)$ 

```
shapiro.test(residuals(lm_poly1)) # Normality test
```

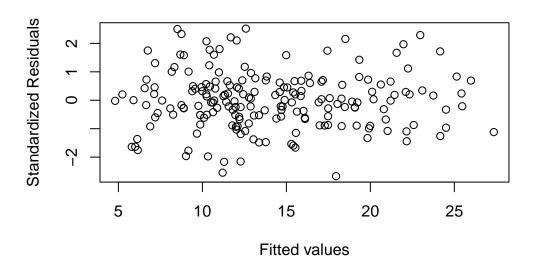
Shapiro-Wilk normality test

```
data: residuals(lm_poly1)
W = 0.98932, p-value = 0.1537
```

```
# Standardized residuals
std_residuals <- rstandard(lm_poly1)
outliers <- which(abs(std_residuals) > 3)
cat("Outliers are at rows:", outliers, "\n")
```

Outliers are at rows:

### Standardized Residuals vs Fitted



```
# --- 12. Model Comparison Table ---
models <- list(
   "Full Model" = lm1,
   "Polynomial Model" = lm_poly,
   "Interaction Model" = lm_interaction,
   "Reduced Polynomial" = lm_poly1
)

aic_table <- data.frame(
   Model = names(models),
   Adjusted_R2 = sapply(models, function(m) round(summary(m)$adj.r.squared, 4)),
   Residual_Std_Error = sapply(models, function(m) round(summary(m)$sigma, 3)),
   AIC = sapply(models, function(m) round(AIC(m), 2))
)
print(aic_table)</pre>
```

```
Model Adjusted_R2 Residual_Std_Error
                                                                          AIC
Full Model
                           Full Model
                                            0.9055
                                                                 1.558 732.33
Polynomial Model
                     Polynomial Model
                                            0.9958
                                                                 0.327 132.62
Interaction Model
                    Interaction Model
                                            0.9763
                                                                 0.780 466.48
Reduced Polynomial Reduced Polynomial
                                            0.9959
                                                                 0.325 125.44
```

```
# --- 13. Cross-Validation (Train/Test Split) ---
set.seed(213)
train_index <- createDataPartition(adver$Sales, p = 0.7, list = FALSE)</pre>
```

Train RMSE: 0.3122478 Test RMSE: 0.3338593

Linear Model RMSE: 0.3187166

```
cat("Random Forest RMSE:", sqrt(mean((rf_pred - adver$Sales)^2)), "\n")
```

Random Forest RMSE: 0.6452739

Best lambda: 0.001244067

```
# Final LASSO model
lasso_model <- glmnet(x, y, alpha = 1, lambda = lasso_cv$lambda.min)
lasso_predictions <- predict(lasso_model, newx = x)
head(lasso_predictions)</pre>
```

s0

1 21.468248

2 10.798021

3 8.652071

4 18.525758

5 13.474984

6 7.892222