

$$Y \sim X_1 + X_2 + X_3 + X_4$$

$$\uparrow X_1 \quad \uparrow Y \Rightarrow \hat{Y} = \hat{\beta}_0 + \hat{\beta}_1 X_1 + \hat{\beta}_2 X_2 + \hat{\beta}_3 X_3 + \hat{\beta}_4 X_4$$

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	-54.28225	7.39453	-7.341	5.41e-08	***
aptitude	0.32356	0.06778	4.774	5.15e-05	***
tol	0.03337	0.07124	0.468	0.6431	
technical	1.09547	0.18138	6.039	1.65e-06	***
general	0.53683	0.15840	3.389	0.0021	**

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 3.549 on 28 degrees of freedom

Multiple R-squared: 0.8768, Adjusted R-squared: 0.8592

F-statistic: 49.81 on 4 and 28 DF, p-value: 2.467e-12

$$t_{calc} = \frac{\hat{\beta} - \beta_0}{S_{\beta}} \quad \downarrow$$

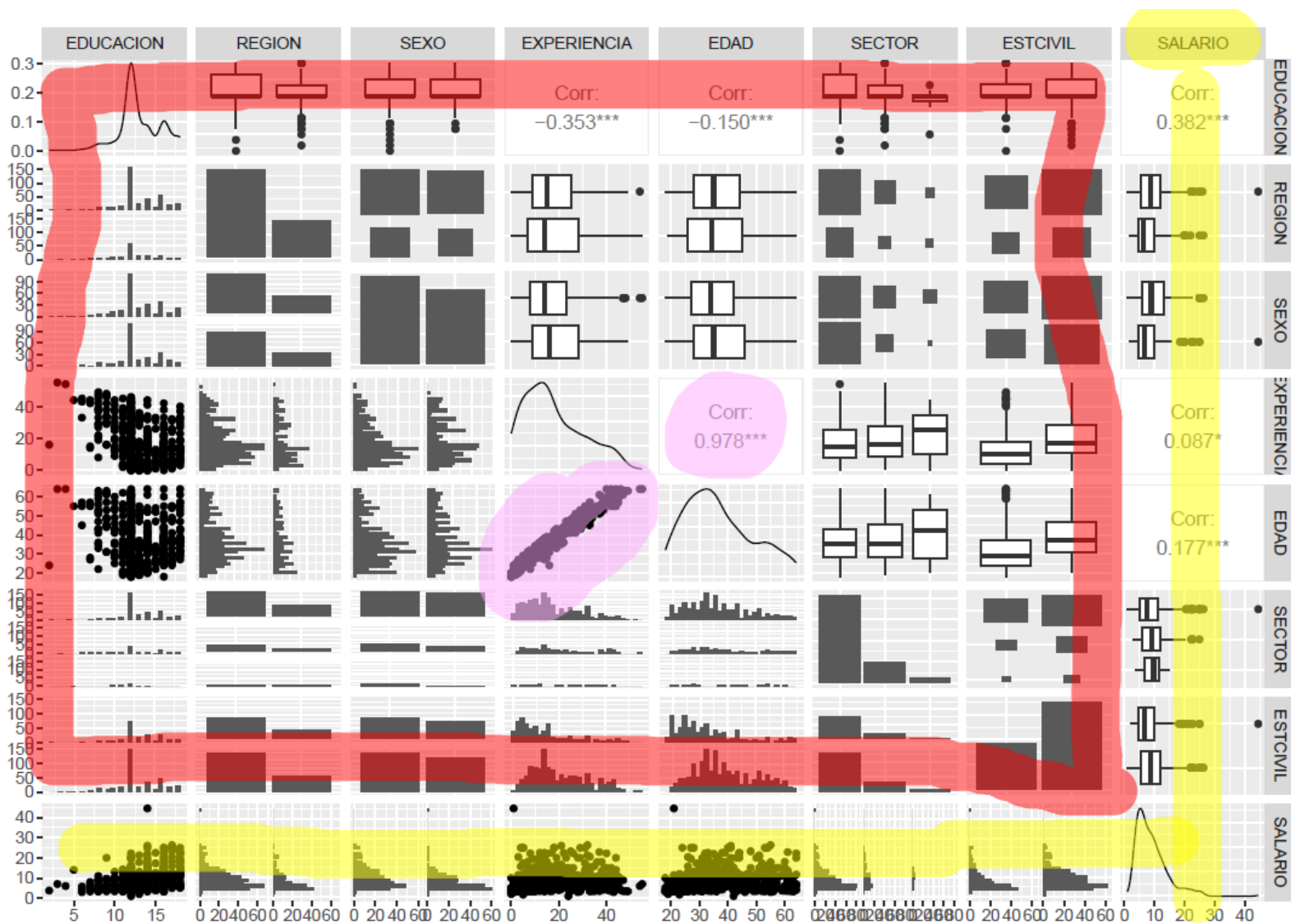
Factor de inflación de varianza (VIF): El VIF de la j -ésima variable viene dado por el j -ésimo término de la diagonal de $(\mathbf{X}'\mathbf{X})^{-1}$. También se puede hallar como $VIF_j = \frac{1}{1-R_j^2}$ donde R_j^2 es el coeficiente de determinación usando la j -ésima variable predictora como respuesta, y las demás predictoras como independientes. Su valor debe ser cercano a 1. Si es mayor a 5 es indicador de multicolinealidad. Si es superior a 10, esta multicolinealidad es muy fuerte. Note la relación entre R^2 y VIF .

$$Y \sim X_1 + X_2 + X_3 + X_4$$

$$VIF_1 = \frac{1}{1-R_1^2} \quad \text{donde } R_1^2 \text{ es el coef. de determinación de } X_1 \sim X_2 + X_3 + X_4 \quad \text{Ideal: } R_1^2 \Rightarrow \text{bajo}$$

$\Rightarrow VIF \rightarrow 1$

$$VIF_2 = \frac{1}{1-R_2^2} \quad \text{donde } R_2^2 \text{ es el coef. de determinación de } X_2 \sim X_1 + X_3 + X_4 \quad \text{Ideal: } R_2^2 \Rightarrow \text{bajo}$$



```
> modelo1 <- ols_step_backward_p(prem = 0.10)
```

Stepwise Summary

Step	Variable	AIC	SBC	SBIC	R2	Adj. R2
0	Full Model	3116.178	3158.982	1599.025	0.26778	0.25662
1	EDAD	3114.310	3152.833	1597.119	0.26760	0.25785

Final Model Output

Model Summary

R	0.517	RMSE	4.394
R-Squared	0.268	MSE	19.307
Adj. R-Squared	0.258	Coef. Var	49.060
Pred R-Squared	0.246	AIC	3114.310
MAE	3.143	SBC	3152.833

RMSE: Root Mean Square Error

MSE: Mean Square Error

MAE: Mean Absolute Error

AIC: Akaike Information Criteria

SBC: Schwarz Bayesian Criteria

ANOVA

	Sum of Squares	DF	Mean Square	F	Sig.
Regression	3766.888	7	538.127	27.455	0.0000
Residual	10309.811	526	19.600		
Total	14076.699	533			

Parameter Estimates

model	Beta	Std. Error	Std. Beta	t	Sig.	lower	upper
(Intercept)	-4.249	1.257		-3.381	0.001	-6.717	-1.780
EDUCACION	0.932	0.081	0.474	11.568	0.000	0.774	1.090
REGION1	-0.839	0.428	-0.074	-1.959	0.051	-1.679	0.002
SEX01	-2.250	0.393	-0.218	-5.727	0.000	-3.022	-1.478
EXPERIENCIA	0.104	0.017	0.250	5.982	0.000	0.070	0.138
SECTOR1	1.024	0.504	0.078	2.032	0.043	0.034	2.014
SECTOR2	0.614	0.957	0.025	0.642	0.521	-1.266	2.495
ESTCIVIL1	0.531	0.420	0.049	1.264	0.207	-0.294	1.356

```
modelo1 <- lm(SALARIO ~ EDUCACION + REGION + SEXO +
EXPERIENCIA + SECTOR + ESTADO CIVIL)
```

```
> modelo |> ols_step_backward_aic()
```

Stepwise Summary						
Step	Variable	AIC	SBC	SBIC	R2	Adj. R2
0	Full Model	3116.178	3158.982	1599.025	0.26778	0.25662
1	EDAD	3114.310	3152.833	1597.096	0.26760	0.25785
2	ESTCIVIL	3113.929	3148.172	1596.662	0.26537	0.25701

Final Model Output

Model Summary			
R	0.515	RMSE	4.401
R-Squared	0.265	MSE	19.365
Adj. R-Squared	0.257	Coef. var	49.088
Pred R-Squared	0.247	AIC	3113.929
MAE	3.154	SBC	3148.172

RMSE: Root Mean Square Error
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ANOVA					
	Sum of Squares	DF	Mean Square	F	Sig.
Regression	3735.570	6	622.595	31.728	0.0000
Residual	10341.129	527	19.623		
Total	14076.699	533			

Parameter Estimates							
model	Beta	Std. Error	Std. Beta	t	sig	lower	upper
(Intercept)	-4.113	1.253		-3.283	0.001	-6.574	-1.652
EDUCACION	0.940	0.080	0.478	11.687	0.000	0.782	1.098
REGION1	-0.828	0.428	-0.073	-1.933	0.054	-1.669	0.013
SEX01	-2.252	0.393	-0.219	-5.730	0.000	-3.025	-1.480
EXPERIENCIA	0.110	0.017	0.264	6.578	0.000	0.077	0.143
SECTOR1	1.041	0.504	0.079	2.066	0.039	0.051	2.032
SECTOR2	0.650	0.957	0.026	0.679	0.497	-1.230	2.531

```
modelo2 <- lm(SALARIO ~ EDUCACION + REGION + SEXO +
EXPERIENCIA + SECTOR)
```

```
> modelo |> ols_step_forward_p(prem = 0.10)
```

Step	Variable	AIC	SBC	SBIC	R2	Adj. R2
0	Base Model	3266.610	3275.171	1750.543	0.00000	0.00000
1	EDUCACION	3184.417	3197.258	1668.446	0.14586	0.14426
2	EXPERIENCIA	3150.099	3167.220	1634.263	0.20202	0.19902
3	SEXO	3116.735	3138.137	1601.287	0.25316	0.24893
4	REGION	3114.403	3140.086	1599.030	0.25919	0.25359
5	SECTOR	3113.929	3148.172	1596.668	0.26537	0.25701
6	ESTCIVIL	3114.310	3152.833	1597.119	0.26760	0.25785

Final Model output

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(Intercept)	-4.249	1.257		-3.381	0.001	-6.717	-1.780
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SEXO1	-2.250	0.393	-0.218	-5.727	0.000	-3.022	-1.478
REGION1	-0.839	0.428	-0.074	-1.959	0.051	-1.679	0.002
SECTOR1	1.024	0.504	0.078	2.032	0.043	0.034	2.014
SECTOR2	0.614	0.957	0.025	0.642	0.521	-1.266	2.495
ESTCIVIL1	0.531	0.420	0.049	1.264	0.207	-0.294	1.356

```
modelo3 <- lm(SALARIO ~ EDUCACION + REGION + SEXO +  
EXPERIENCIA + SECTOR + ESTCIVIL) == modelo1
```

```
> modelo |> ols_step_forward_aic()
```

Step	variable	AIC	SBC	SBIC	R2	Adj. R2
0	Base Model	3266.610	3275.171	1750.543	0.00000	0.00000
1	EDUCACION	3184.417	3197.258	1668.446	0.14586	0.14426
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5	SECTOR	3113.929	3148.172	1596.668	0.26537	0.25701

Final Model output

R	0.515	RMSE	4.401
R-Squared	0.265	MSE	19.365
Adj. R-Squared	0.257	Coef. Var	49.088
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RMSE: Root Mean Square Error
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	Sum of Squares	DF	Mean Square	F	sig.
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model	Beta	Std. Error	Std. Beta	t	sig.	lower	upper
(Intercept)	-4.113	1.253		-3.283	0.001	-6.574	-1.652
EDUCACION	0.940	0.080	0.478	11.687	0.000	0.782	1.098
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SECTOR1	1.041	0.504	0.079	2.066	0.039	0.051	2.032
SECTOR2	0.650	0.957	0.026	0.679	0.497	-1.230	2.531

```
modelo4 <- lm(SALARIO ~ EDUCACION + REGION + SEXO +  
EXPERIENCIA + SECTOR) == modelo2
```

```
# A tibble: 9 x 5
```

	term <chr>	estimate <dbl>	std.error <dbl>	statistic <dbl>	p.value <dbl>
1	(Intercept)	-1.84	6.80	-0.271	0.787
2	EDUCACION	1.33	1.12	1.19	0.233
3	REGION1	-0.835	0.429	-1.95	0.0519
4	SEX01	-2.26	0.394	-5.73	0.0000000166
5	EXPERIENCIA	0.505	1.12	0.453	0.651
6	EDAD	-0.401	1.11	-0.360	0.719
7	SECTOR1	1.03	0.504	2.03	0.0424
8	SECTOR2	0.611	0.958	0.638	0.524
9	ESTCIVIL1	0.538	0.421	1.28	0.202

Hay multicolinealidad

```
modelo2 |> tidy()
```

```
# A tibble: 8 x 5
```

	term <chr>	estimate <dbl>	std.error <dbl>	statistic <dbl>	p.value <dbl>
1	(Intercept)	-4.87	1.31	-3.72	2.22e- 4
2	EDUCACION	0.829	0.0763	10.9	6.19e-25
3	REGION1	-0.840	0.428	-1.96	5.04e- 2
4	SEX01	-2.25	0.393	-5.72	1.77e- 8
5	EDAD	0.104	0.0173	5.98	4.23e- 9
6	SECTOR1	1.02	0.504	2.03	4.27e- 2
7	SECTOR2	0.616	0.957	0.643	5.20e- 1
8	ESTCIVIL1	0.530	0.420	1.26	2.08e- 1