# Reproducible Research Template $^*$

EB

## Preliminary – please do not quote

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#### Abstract

Some abstract here.

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### 1 Introduction

If you are using this template, please cite this item from the references: von Gaudecker (2023).

The data set for the example project is taken from https://www.stem.org.uk/resources/elibrary/resource/28452/large-datasets-stats4schools. It contains data on smoking habits in the UK, with 1691 observations and 12 variables. We consider only 4 of the 12 features for the prediction of the variable smoking: marital\_status, highest\_qualification, gender and age. We model the dependence using a Logistic model. All numerical features are included linearly, while categorical features are expanded into dummy variables. Figures below illustrate the model predictions over the lifetime. You will find one figure and one estimation summary table for each installed programming language.

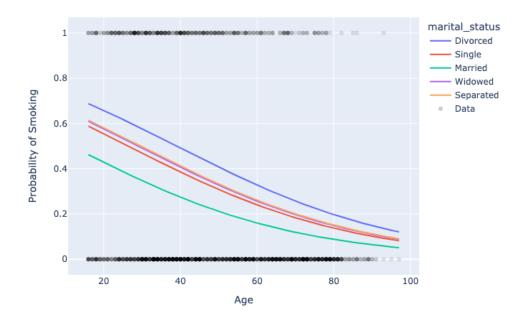


Figure 1: *Python:* Model predictions of the smoking probability over the lifetime. Each colored line represents a case where marital status is fixed to one of the values present in the data set.

#### References

Gaudecker, Hans-Martin von (2023). "Templates for Reproducible Research Projects in Economics". https://doi.org/10.5281/zenodo.7780520.

Dep. Variable:	smoke num	orical I	No Obser	rvations	: 16	01	
•	Model: Smoke_num		No. Observations: Df Residuals:			1677	
Method: MLE			Df Model:			13	
Date:	Thu, 25 Jan		Pseudo R	-	0.08683		
Time:	10:03:46		Log-Likeli	hood:	-866.58		
${f converged}$ :	True		LL-Null:		-948.98		
Covariance Type	nonrobust		LR p-val	lue:	2.103e-28		
		coef	std err	$\mathbf{z}$	$P> \mathbf{z} $	[0.025]	0.975]
Intercept		0.3872	0.363	1.068	0.286	-0.324	1.098
${f gender}[{f T.Male}]$		0.1731	0.122	1.417	0.157	-0.066	0.413
${ m marital\_status}[{ m T.Married}]$		-0.9404	0.196	-4.806	0.000	-1.324	-0.557
$\operatorname{marital\_status}[\operatorname{T.Separated}]$		-0.3234	0.318	-1.019	0.308	-0.946	0.299
$\operatorname{marital\_status[T.Single]}$		-0.4332	0.216	-2.008	0.045	-0.856	-0.010
${ m marital\_status}[{ m T.Widowed}]$		-0.3441	0.262	-1.312	0.189	-0.858	0.170
$\overline{\text{qualification}}[ ext{T.Degree}]$		-0.1773	0.309	-0.574	0.566	-0.783	0.428
${f qualification [T.GCSE/CSE]}$		0.8614	0.330	2.609	0.009	0.214	1.508
${\bf qualification[T.GCSE/O\ Level]}$		0.8233	0.283	2.909	0.004	0.269	1.378
${\bf qualification [T. Higher/Sub\ Degree]}$		0.4487	0.340	1.318	0.187	-0.218	1.116
qualification[T.No Qualification]		0.9490	0.288	3.297	0.001	0.385	1.513
${\bf qualification[T.ONC/BTEC]}$		0.5820	0.362	1.607	0.108	-0.128	1.292
${\bf qualification [T.Other/Sub~Degree]}$		0.8425	0.339	2.485	0.013	0.178	1.507
age		-0.0342	0.005	-7.179	0.000	-0.044	-0.025

Table 1: Python: Estimation results of the linear Logistic regression.