

Appendix A: Samples of Datasets Used in Empirical Study

Figure 6 Snippet of Dataset 1.

	sl_no	gender	ssc_p	ssc_b	hsc_p	hsc_b	hsc_s	degree_p	degree_t	workex	etest_p	specialisation	mba_p	status	salary
0	1	M	67.00	Others	91.00	Others	Commerce	58.00	Sci&Tech	No	55.0	Mkt&HR	58.80	Placed	270000.0
1	2	M	79.33	Central	78.33	Others	Science	77.48	Sci&Tech	Yes	86.5	Mkt&Fin	66.28	Placed	200000.0
2	3	M	65.00	Central	68.00	Central	Arts	64.00	Comm&Mgmt	No	75.0	Mkt&Fin	57.80	Placed	250000.0
3	4	M	56.00	Central	52.00	Central	Science	52.00	Sci&Tech	No	66.0	Mkt&HR	59.43	Not Placed	NaN
4	5	M	85.80	Central	73.60	Central	Commerce	73.30	Comm&Mgmt	No	96.8	Mkt&Fin	55.50	Placed	425000.0

Figure 7 Snippet of Dataset 2.

	enrollee_id	city	city_development_index	gender	relevent_experience	enrolled_university	education_level	major_discipline	experience	company_size	company_type	last_new_job	training_hours	target
0	8949	city_103	0.920	Male	Has relevent experience	no_enrollment	Graduate	STEM	>20	NaN	NaN	1	36	1.0
1	29725	city_40	0.776	Male	No relevent experience	no_enrollment	Graduate	STEM	15	50-99	Pvt Ltd	>4	47	0.0
2	11561	city_21	0.624	NaN	No relevent experience	Full time course	Graduate	STEM	5	NaN	NaN	never	83	0.0
3	33241	city_115	0.789	NaN	No relevent experience	NaN	Graduate	Business Degree	<1	NaN	Pvt Ltd	never	52	1.0
4	666	city_162	0.767	Male	Has relevent experience	no_enrollment	Masters	STEM	>20	50-99	Funded Startup	4	8	0.0

Appendix B: Datasets 1 and 2: Detailed Description

Table 9 Dataset 1

Column Name	Data Type	Range of Values	Missing Values (%)
sl_no	int64	1 to 215	0
gender	object	M, F	0
ssc_p	float64	40.89 to 89.4	0
ssc_b	object	Central, Others	0
hsc_p	float64	37 to 97.7	0
hsc_b	object	Central, Others	0
hsc_s	object	Commerce, Science, Arts	0
degree_p	float64	50 to 91	0
degree_t	object	Comm&Mgmt, Sci&Tech, Others	0
workex	object	Yes, No	0
etest_p	float64	50 to 98	0
specialisation	object	Mkt&Fin, Mkt&HR	0
mba_p	float64	51.21 to 77.89	0
status	object	Placed, Not Placed	0
salary	float64	20,000 to 94,000	31.2

Table 10 Dataset 2

Column Name	Data Type	Range of Values	Missing Values (%)
enrollee_id	int64	1 to 33380	0
city	object	City_1, City_2, ... City_123	0
city_development_index	float64	0.448 to 0.949	0
gender	object	Male, Female, Other	23.5
relevant_experience	object	No relevant experience, Has relevant experience	0
enrolled_university	object	no_enrollment, Part time course, Full time course	2
education_level	object	Primary School, High School, Graduate, Masters, Phd	2.4
major_discipline	object	No Major, Arts, Business Degree, Humanities, STEM, Other	14.7
experience	object	<1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17, 18, 19, 20, >20	0.3%
company_size	object	<10, 10/49, 50-99, 100-500, 500-999, 1000-4999, 5000-9999, 10000+	31
company_type	object	Pvt Ltd, Funded Startup, Public Sector, Early Stage Startup, NGO, Other	32
last_new_job	object	never, 1, 2, 3, 4, >4	2.2
training_hours	int64	1 to 336	0
target	float64	0, 1	0

Appendix C: List of Datasets and Data Wrangling Recipes

Table 11 Dataset 1

Source Link: <https://www.kaggle.com/datasets/benroshan/factors-affecting-campus-placement>.

Date Accessed: 3rd February 2023

No.	Link
1	https://www.kaggle.com/code/benroshan/you-re-hired-analysis-on-campus-recruitment-data
2	https://www.kaggle.com/code/priyankarao18/eda-visualization
3	https://www.kaggle.com/code/arindambaruah/placement-predictions-using-log-reg-knn-rfc-xgb
4	https://www.kaggle.com/code/avaniudupa/campusrecruitment-predict-if-a-student-gets-placed
5	https://www.kaggle.com/code/achintyatripathi/highcharts-eda-logistic-reg-prediction
6	https://www.kaggle.com/code/adityaaggarwal09/a-bull-s-eye-on-prediction
7	https://www.kaggle.com/code/poojaravi01/campus-placement-prediction-model
8	https://www.kaggle.com/code/ami4501/kernel3fdc106f66
9	https://www.kaggle.com/code/mani97/placed-or-not-eda-classification-88-8
10	https://www.kaggle.com/code/saitej31/campus-recruitment-90-acc-logistic-regression

Table 12 Dataset 2

Source Link:

<https://www.kaggle.com/datasets/arashnic/hr-analytics-job-change-of-data-scientists>.

Date Accessed: 3rd February 2023

No.	Link
1	https://www.kaggle.com/code/josephchan524/hranalytics-lightgbm-classifier-auc-80
2	https://www.kaggle.com/code/gokulrajkmv/hr-analytics-attribution-prediction
3	https://www.kaggle.com/code/elyousfiomar/hyperparameter-tuning-gradient-boosting
4	https://www.kaggle.com/code/gdeepakkumar/looking-for-a-job
5	https://www.kaggle.com/code/stuartday274/job-change-predictions-using-a-pipeline
6	https://www.kaggle.com/code/ihormuliar/hr-analytics
7	https://www.kaggle.com/code/italomarcelo/predictions-hr-job
8	https://www.kaggle.com/code/anmol686/job-change-prediction
9	https://www.kaggle.com/code/harshbhatnagar/hr-analystics-prediction
10	https://www.kaggle.com/code/nicklcorona/predict-data-science-job-change
11	https://www.kaggle.com/code/scharom/old-but-interpretable-the-logit-model
12	https://www.kaggle.com/code/gcmadhan/hr-analysis-data-scientist-job
13	https://www.kaggle.com/code/raj5kumar5/hr-analytics-job-change
14	https://www.kaggle.com/code/artiomkolas/hr-analytics
15	https://www.kaggle.com/code/meet1415/beginner-notebook-for-hr-analytics
16	https://www.kaggle.com/code/ouyangwenguang/hr-analytics-job-change-of-data-scientists
17	https://www.kaggle.com/code/naiara/hr-analytics-job
18	https://www.kaggle.com/code/adrianabellanoza/aabellanoza-hr-analytics-job-change
19	https://www.kaggle.com/code/kfhuang/hr-analytics
20	https://www.kaggle.com/code/kowalsky999/lightgbm-predict-hr-analytics
21	https://www.kaggle.com/code/bashtovoy/notebooke3cc6c23a8
22	https://www.kaggle.com/code/navesammy/hr-analytics
23	https://www.kaggle.com/code/nicolamignoni/job-change-of-data-scientists