

# Lead Scoring Case Study

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### Problem Statement



- X Education, an education company that sells online courses to industry professionals, has appointed us to help them select the most promising leads.
- The company requires us to build a model wherein we need to assign a lead score between 0 and 100 to each of the leads which can be used by the company to target potential leads. A higher score would mean that the lead is hot, i.e. is most likely to convert whereas a lower score would mean that the lead is cold and will mostly not get converted.
- The CEO has given a ballpark of the target lead conversion rate to be around 80% (currently at 30%).
- There are some more problems presented by the company which our model should be able to adjust to if the company's requirement changes in the future so you will need to handle these as well.

### Data Preparation

Dropped columns with large percent of null values

Dropped categorical columns with only all or most values of single category

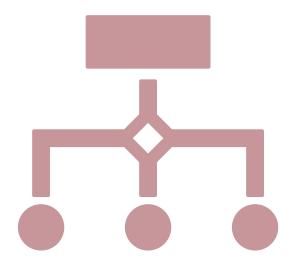
Dropped unnecessary columns such as Prospect ID, Last Activity, Country, Specialization etc.

Converted binary variables Yes/ No to 0/1 for conversion to numerical variables

Converted categorical variables to dummy variables

Handled outliers in columns by capping till 99.5th percentile

Performed test train split and scaled features



# Model Building and Validation

- Used statsmodel to create the first logistic regression model on the train data set and dropped irrelevant columns
- •Used Recursive feature elimination (RFE) to select the 7 most important features
- Checked VIF of features to eliminate multicollinearity
- Assessed the model with remaining columns and removed any columns with high P values
- Utilized the precision-recall curve to find out the optimal cut-off for lead score for required precision value of 80 percent.
- Checked the precision value of the model against test data set to check if it meets the required criteria.

## Final Model

Generalized Linear Model Regression Results							
Dep. Variable:	Converted		No. Observations:		s:	6339	
Model:	GLM		Df Residuals		: 6332		
Model Family:	Binomial		Df Model		6		
Link Function:	Logit		Scale:		e: 1.	.0000	
Method:	IRLS		Log-Likelihood:		d: -2	884.9	
Date:	Sun, 15 De	c 2024	<b>Deviance:</b> 5769.8			769.8	
Time:	13	3:54:40	Pearson chi2: 6.41e+03			e+03	
No. Iterations:		6 <b>F</b>	Pseudo R	-squ. (CS	<b>):</b> 0	.3424	
Covariance Type:	nonrobust						
		coef	std err	z	P> z	[0.025	0.975]
	const	-0.3170	0.057	-5.524	0.000	-0.429	-0.204
Do Not Email		-1.4293	0.162	-8.803	0.000	-1.748	-1.111
Total Time Spent of	on Website	0.9829	0.035	28.398	0.000	0.915	1.051
Landing Page S	ubmission	-0.3908	0.071	-5.530	0.000	-0.529	-0.252
Lead	Add Form	3.7831	0.218	17.367	0.000	3.356	4.210
Unknown Employment		-1.2578	0.082	-15.346	0.000	-1.418	-1.097
Working Pr	ofessional	2.4713	0.179	13.776	0.000	2.120	2.823

#### Recommendations

As we can see from the final model, Lead Origin (Lead Add Form), What is your current occupation (Working Professional, Unknown Employment) and Do Not Email are top three variables which contribute towards the probability of a lead getting converted as they have the highest absolute value of coefficients in the model.



As per the final model shown earlier, below 3 categorical/dummy variables should be focused upon to increase the probability of lead conversion-

Lead Add Form- Leads originated via Lead Add form have the highest probability of conversion

Working Professional- Leads with employment status as Working Professional

Do Not Email- Leads which have selected Do No Email would most likely not avail the course and hence should not be focused upon

### Recommendations



Strategy to increase conversion by additional interns: Since we are having additional resources in our team in form of interns, we suggest lowering the threshold of lead score from the existing value of 60 (for the required conversion rate of 80%). Since, as per the precision-recall curve, optimum cut off point is 40, we suggest taking the cut-off point as 40 to increase the number of probable leads we can call.



Strategy to minimize useless phone calls during quarter end: Given we need to reduce the number of useless phone calls, we need to maximize the conversion rate i.e. precision value for this case. We suggest taking the cut-off point for lead score as 80 which would result in precision value of 88% and hence result in only c. 12% incorrect calls.