

Section	Ask	Points	What good looks like	What average looks like	What poor looks like	What nothing looks like	Weightage
		60	80-100%	60-80%	<60%	0	100.00%
Exploratory Data Analysis	<ul style="list-style-type: none"> - Problem definition - Univariate analysis - Bivariate analysis - Use appropriate visualizations to identify the patterns and insights - Key meaningful observations on individual variables and the relationship between variables 	8	<p>1) Defined problem statement, Observations on shape of data, data types of various attributes, statistical summary. (2)</p> <p>2) Univariate and Bivariate analysis (variable distributions, interactions between variables) to understand the relationships in data beyond the set of questions already provided (2)</p> <p>3) The analysis includes answers for Insight-based questions asked (2)</p> <p>Illustrate the insights based on EDA: (2)</p> <p>1) Commented on range of attributes, outliers of various attributes.</p> <p>2) Commented on the distribution of the variables and potential fixes for better modelling.</p> <p>3) Key insights shared from univariate and bivariate analysis.</p>	<p>1) Definition of problem, Observations on shape of data, data types of various attributes, statistical summary.</p> <p>2) The analysis includes answers for Insight-based questions.</p> <p>Illustrate the insights based on EDA:</p> <p>1) Commented on range of attributes, outliers of various attributes</p> <p>2) Insights provided are very brief or missing for plots in univariate and bivariate analysis</p>	<p>1) Definition of problem (as per given problem statements) and Observations on data types of various attributes, statistical summary.</p> <p>Illustrate the insights based on EDA:</p> <p>1 - Univariates done with no comments</p> <p>2 - A few random bivariate done with little commentary</p>	<p>1) No EDA is done</p> <p>2) Did not comment on range of attributes, outliers of various attributes</p> <p>3) Did not provide insights for any plot</p>	13.33%
Data Preprocessing	<ul style="list-style-type: none"> - Prepare the data for analysis - Missing Value Detection and Treatment (if needed with rationale) - Outlier Detection and Treatment (if needed with rationale) - Feature Engineering (if needed with rationale) - Prepare data for modeling 	5	<p>1) Dropped case_id column with a reason for the same. (1)</p> <p>2) Identified and treated the negative values in number of employees column. (2)</p> <p>3) Identified outliers and provided comments on it - Treatment is not necessary but a valid reasoning should be provided for treatment/non-treatment. (1)</p> <p>4) Prepared the data by dividing it into train, test, and validation sets (1)</p>	<p>1) Dropped case_id column with a reason for the same.</p> <p>2) Identified and treated the negative values in number of employees column.</p> <p>3) Identified outliers but no comments are provided</p>	<p>1) Dropped case_id column with a reason for the same.</p> <p>2) Identified and treated the negative values in number of employees column.</p>	No data pre-processing done	8.33%
Model Building - Original Data	<ul style="list-style-type: none"> - Build atleast 5 classification models (Using decision trees, random forest, bagging classifier and boosting methods) * You can choose not to build XGBoost if you are facing issues with installation 	6	<p>1) Choose the metric of choice with proper rationale [1]</p> <p>2) Build 5 classification models and train them using the original train data [1x5]</p>	<p>1) Build 3-5 classification models and trained them using the original train data</p>	<p>1) Build < 3 classification models and trained them using the original train data</p>	<p>1) Did not train models with default hyperparameters or</p> <p>2) Trained regressors instead of classifiers or</p> <p>3) Did not run the cell/ result is an error</p>	10.00%
Model Building - Oversampled Data	<ul style="list-style-type: none"> - Build atleast 5 classification models using oversampled train data (Using decision trees, random forest, bagging classifier and boosting methods) * You can choose not to build XGBoost if you are facing issues with the installation 	6	<p>1) Oversample the train data [1]</p> <p>2) Build 5 classification models and train them using oversampled data [1x5]</p>	<p>1) Oversampled the train data</p> <p>2) Build 3-5 classification models and trained them using oversampled data</p>	<p>1) Oversampled the train data</p> <p>2) Build less than 3 classification models and trained them using oversampled data</p>	<p>1) Did not train models with default hyperparameters or</p> <p>2) Trained regressors instead of classifiers or</p> <p>3) Did not run the cell/ result is an error</p> <p>4) Did not choose any scorer (or chose accuracy) to tune the models</p>	10.00%

Model Building - Undersampled Data	<ul style="list-style-type: none"> - Build atleast 5 classification models using undersampled train data (Using decision trees, random forest, bagging classifier and boosting methods) * You can choose not to build XGBoost if you are facing issues with the installation 	6	1) Undersample the train data [1] 2) Build 5 classification models and train them using undersampled data [1x5]	1) Undersampled the train data 2) Build 3-5 classification models and trained them using undersampled data	1) Undersampled the train data 2) Build less than 3 classification models and trained them using undersampled data	1) Did not train models with default hyperparameters or 2) Trained regressors instead of classifiers or 3) Did not run the cell/ result is an error	10.00%
Hyperparameter Tuning	<ul style="list-style-type: none"> - Choose atleast 3 best performing models among all the models built previously (Mention the reason for the choices made) - Tune the chosen models - Check the performance of the tuned models 	10	1) Choose atleast 3 best performing models among all the models built previously and mention proper reasoning for choosing those models [1] 3) Tune the chosen models with metric of interest [2.5x3] 4) Check the performance of the tuned models [0.5x3]	1) Choose atleast 3 models 2) Tune the chosen models	1) Choose and tune less than 3 models obtained above with metric of interest	1) Did not train models with default hyperparameters or 2) Trained regressors instead of classifiers or 3) Did not run the cell/ result is an error 4) Did not choose any scorer (or chose accuracy) to tune the models	16.67%
Model Performances	<ul style="list-style-type: none"> - Compare performances of the tuned models and choose a final model. - Check the performance of final model on test data. 	5	1) Compare model performances of tuned models [2] 2) Choose best model [1] 3) Find test performance [1] 4) Metric of choice > 0.8 on the test set [1]	1) Compare model performances of tuned models 2) Choose best model 3) Find test performance 4) 0.75 < Metric of choice < 0.8 on test data	1) Choose best model 2) Metric of choice < 0.75 on test data		8.33%
Actionable Insights & Recommendations	<ul style="list-style-type: none"> - Compare model performance on various metrics. - Conclude with the key takeaways for the business 	6	1) 3-4 Conclusions provided on the model performance comparison and EDA, key takeaways in form of important features identified. (4) 2) 2-3 Recommendations mentioned (2) [Recommendations can also include points on additional data sources for further analysis, model implementation in real world, potential business benefits from improving the model, etc.]	1) 1-2 Conclusions provided on the model performance comparison and EDA, key takeaways in form of important features identified. 2) 1 Recommendation mentioned	1) Any Conclusions provided on the model performance comparison and EDA 2) No recommendations	None of the steps are performed.	10.00%

Presentation / Notebook - Overall quality	<ul style="list-style-type: none"> - Structure and flow - Crispness - Visual appeal - Conclusion and Business Recommendations <p>OR</p> <ul style="list-style-type: none"> - Structure and flow - Well commented code - Conclusion and Business Recommendations 	8	<ul style="list-style-type: none"> 1) Clear structure and flow - everything sits well in a story 2) Crispness (Not too many words, just enough to keep the focus on key things/points) 3) Visual appeal (Use of charts, colors, diagrams, format, symmetry, informative visualizations that are easy to interpret) <p>OR</p> <ul style="list-style-type: none"> 1) Well structured notebook with a logical flow 2) Clean and well commented code 	<ul style="list-style-type: none"> 1) There is structure and flow but some bits are jumbled 2) Points are made but in too many words 3) Lesser charts, format is not the cleanest <p>OR</p> <ul style="list-style-type: none"> 1) There is structure and flow but some bits are missing 2) Some of the code is commented <p>* If any section is missing and points have been deducted for the same in a previous section, then no points should be deducted in this section</p>	<ul style="list-style-type: none"> 1) No structure or flow 2) Only a few points are covered, story is not complete 3) Not many visuals used <p>OR</p> <ul style="list-style-type: none"> 1) no structure or flow 2) no comments in the code 	<p>No report</p> <p>OR</p> <p>No code</p>	13.33%
--	--	----------	---	--	--	---	---------------