MANAS BHAT



Year	MIC QUALIF	ICATIONS		
- cui		Degree /Board	University /Institution	%/CGPA
2025*	Post Gradua	te Diploma in Business Analytics	IIM Calcutta, IIT Kharagpur, ISI Kolkata	-
2022		S. (Dual Degree) Mathematics	Indian Institute of Science Education and Research Bhopal	7.61/10
2017		CLASS XII	DHSS, Jammu	89.40 %
2015		CLASS X	MHACS Nagbani, Jammu	10/10
	ILLS/TOOLS		al Data Analysis, Critical Thinking, Python, R Programming, SQL,	
	DS AND ACH			
	ters Thesis		Thesis in Pure Math at IISER Bhopal on the topic 'Zeros of Moo	dular Form
Scholastic Achievements		■ Secured AIR 2 in PGDBA entrar ■ Scored 98/122 in Math section	nce exam 99.83 & 99.87 %ile in Quant section of CAT 2022 at of JEE ADVANCED 2017 Secured 99.93 %ile (out of 22.5k) in	nd XAT 202 JK-CET 201
		■ Attended INMO training camp after securing Rank 5 in Regional Mathematics Olympiad (RMO) in 201 ■ Cleared National Standard Examination in Physics(NSEP'15) held by IAPT, with 44k+ candidates enrolled		
Case Competitions		 National Winner among 823 teams in Mind Wizz; a Suppy chain analytics comp. held by IIM Udaipur['2 National Finalist (Top 7/424) in μ-lytics; an HR analytics comp. conducted by MU School of Business['2 		
ACADE	MIC PROJEC	ΓS		
Used Cars Price Prediction (Regression)		 ■ Developed Regression models to predict the price of used cars using 11.8k+ observations and 20 feature ■ Performed feature engineering; imputed missing values using iterative imputer; validated all assumption ■ Tackled multicollinearity with VIF, PCA, Lasso and Ridge; handled influential points using dffits-statist ■ Adj. R² = 0.77 (↑ 11.5%); Analyzed non-linear models: AdaBoost, XGB, RF; enhanced Adj. R² to 0.84 (↑ 9%) 		
Drug-Drug Interaction Prediction		■ Harnessed ~1.92L drug pairs to formulate Drug-Drug interaction classification models using 86 DDI classed Combined drug chemical structures into feature vectors; performed feature selection & used DeepNN, KN ■ Executed stacking ensemble(RF, NN & XGB) with LR as meta learner & achieved accuracy of 0.94 (↑4.5% ■ Explored Graph Neural Networks and node embedding techniques for predicting drug-drug interactions.)		
Online Retail Segmentation		■ Investigated UK online retail dataset containing ~3.5L observations for identifying diff. customer segment Performed feature engg.; Cohort & RFM analysis; utilized t-SNE; used K-means, agglomerative & DBSCA Identified 4 segments with customer profiling from elbow method; silhouette score- 0.83 & DB index- 0.4		
Brand Positioning through Perceptual Map (NLP)		■ Created Perceptual Map of 8 Hotel chains using OpenAI embeddings, BERTopic & LDA from 12k+ review Selected no. of topics by coherence score & Employed LR to describe thematic patterns across hotel chain Leveraged Multidimensional Scaling yielding actionable insights while achieving a stress score of 0.09		
UK Electronics Sales Forecast		■ Forecasted monthly sales of electronics for a period of 30 months from the past 36 years data using SARIM ■ Used ADF test for Stationarity ; log transformed for heteroscedasticity; fixed AR/MA orders by ACF/PAC ■ Utilized AIC , BIC for best model selection; verified residuals using Ljung-Box test; achieved MAPE of 4.33 ·		
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