List of my best posts – 2021

- 1) Things that can get you rejected in a Data Science Interview https://bit.ly/3EHIb1m
- 2) What do recruiters look for while hiring Data Science Interns? https://bit.ly/3eysVI5
- 3) Why saying "We accept the Null Hypothesis" is wrong https://bit.ly/3qqxbiu
- 4) Logistic Regression is Regression (meme) https://bit.ly/3qsHhiV
- 5) The many names of Dependent and Independent variable https://bit.ly/33Uwfv6
- 6) Most Important Assumption Checks of Linear Regression https://bit.ly/3eu47ky
- 7) First principles thinking in Data Science. https://bit.ly/3Fx4Ey2
- 8) Entropy and Getting Hired. https://bit.ly/32Na3CQ
- 9) AutoML dangers: Abstraction and Data Science Not a great combination!! https://bit.ly/33Y84ff
- 10) Fallacy of "Modelling will take only 20% time" https://bit.ly/3sK5W5c
- 11) Be wary of Automated Feature Selection Chi Square test of Independence https://bit.ly/3ENhsPT
- 12) Central Limit theorem does not kick in at n=30 https://bit.ly/3pyYVCa
- 13) Data Science is a team effort https://bit.ly/3px3BbO
- 14) Feature Augmentation https://bit.ly/3z8HDiK
- 15) Lost in translation https://bit.ly/3pxrVKA
- 16) Not knowing how the ML algorithms works under the hood leads to acceptance of 'default' - https://bit.ly/3FulEE1
- 17) At what stage should a Data Scientist be roped in? https://bit.ly/3Jnmo14

- 18) Descriptive statistics & good data visualizations are your two best friends https://bit.ly/3FC6TQC
- 19) Logistic Regression: Classification if you use Python, Regression if you use R!! https://bit.ly/32JdfPu
- 20) Errors and Residuals https://bit.ly/3JjTcbz
- 21) Normal Distribution may be popular distribution but it is not the most prevalent. https://bit.lv/3EvSvYa
- 22) Is correlation not causation? https://bit.ly/3Fz5dHC
- 23) There never was a Normal Distribution https://bit.ly/3ewMa4W
- 24) Linear regression does not model the raw values of Dependent Variable https://bit.ly/3mGmHKN
- 25) AutoML dangers: "Try all models" https://bit.ly/3euOV6C
- 26) Clearing Misconceptions around Confidence Interval. https://bit.ly/3HqOOIn
- 27) The ML algorithm is the nucleus of a Data Science project https://bit.ly/3mEmQOG
- 28) Deep Learning is like Disneyland https://bit.ly/316TQrB
- 29) Real Statisticians / Data Scientists don't test for normality. https://bit.ly/3pzlert
- 30) All models are wrong, some are useful" ≠ "Modeling is a futile exercise -

https://bit.ly/3z6E9wT

- 31) No Full stack Data Scientist https://bit.ly/3sDDr9c
- 32) The problem of learning "Formula First" in Data Science https://bit.ly/3z7kCMV
- 33) Dunning-Kruger Effect in Data Science https://bit.ly/3FB4xS1
- 34) The XY problem in Data Science https://bit.ly/3HmgsUa
- 35) Stock markets can't be predicted https://bit.ly/3sBJ6wF
- 36) There is no Poetic License in Data Science https://bit.ly/3FGrwLB

- 37) β -hat vs Y-hat problem https://bit.ly/3mFqdFq
- 38) Why the word AI https://bit.ly/3sCqY50
- 39) Assumption Selection https://bit.ly/3EBIOZ9
- 40) German Tank Problem https://bit.ly/345BHvm
- 41) Clearing Confusion around Skewness of Probability Distribution https://bit.ly/32EuNwi
- 42) What is the harm in calling Logistic Regression a classification algorithm? https://bit.ly/3Jo0zP2
- 43) Why Confusion matrix is called so? https://bit.ly/32IWLH7
- 44) The curious case of statisticians recommending Cross Validation. https://bit.ly/32EvCVV
- 45) 'Red Pill' moments in Data Science https://bit.ly/3FvVulr
- 46) Things that surprise New Data Scientists when they first step into the corporate world. https://bit.ly/3sEwbt0
- 47) Intuitive Explanations to Data Science Question (Stack exchange) https://bit.ly/3eMGG6p
- 48) "You don't need enough math" Biggest lie told to Data Scientists https://bit.ly/3Hns0qx
- 49) Heuristics to judge Data Science courses https://bit.ly/3zevvwE
- 50) Formula 1, Kaggle and Applied Data Science https://bit.ly/3mE1PnB
- 51) Know Your Data First Time series, Cross Sectional & Panel Data https://bit.ly/3JmFeFH
- 52) Lessons for Data Scientists from Zillow fallout. https://bit.ly/3Ji8U60
- 53) Market Research A unique but efficient path to become a Data Scientist.https://bit.ly/346fRYD
- 54) How Aryma Labs Internship is a little different https://bit.ly/3Ji9hOK

- 55) Putting ML algo in production is akin to putting a jelly on a conveyor belt https://bit.ly/33YI1Wp
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- 61) Parallels Between Data Dredging and Algorithm Dredging. https://bit.ly/3EyBcGD
- 62) You can google how to code but can't google which statistical technique to use https://bit.ly/3quDBNt