



# LOGISTICS REGRESSION

**EMBARKING ON A JOURNEY INTO  
DATA SCIENCE**

**YA MANON**

**You can have data without information but  
you cannot have information without data.**

**-Daniel Keys Maran**

# ABOUT THIS SERIES

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## **PART 1**

Data Prep & EDA



## **PART 2**

Regression

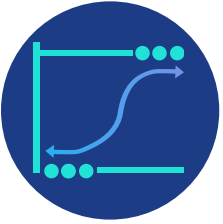


## **PART 3**

Classification

# LOGISTICS REGRESSION

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In this section we'll cover the a technique that predict a categorical dependent variable based on one or more independent variables

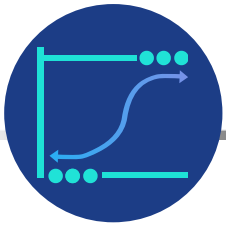
## TOPICS WE'LL COVER:

Logistics Regresson

Goals of Regression

Evaulation

## GOALS FOR THIS SECTION:



# LOGISTICS REGRESSION

**Logistic Regression** is a classification technique used to predict the probability of a binary (true/false) outcome

- In its simplest form, logistic regression forms an **S-shaped curve between 0 -1**, which represents the probability of a TRUE outcome for any given value of X
- The **likelihood function** measures how accurately a model predicts outcomes, and is used to optimize the “shape” of the curve
- Although it has the word “*regression*” in its name, logistic regression is not used for predicting numeric variables

## **Example use cases:**

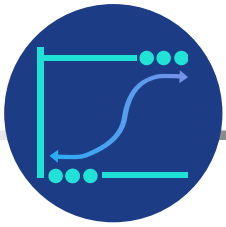
Flagging spam emails or fraudulent credit card transactions

Logistics Regression

Goals of Regression

Evaluation



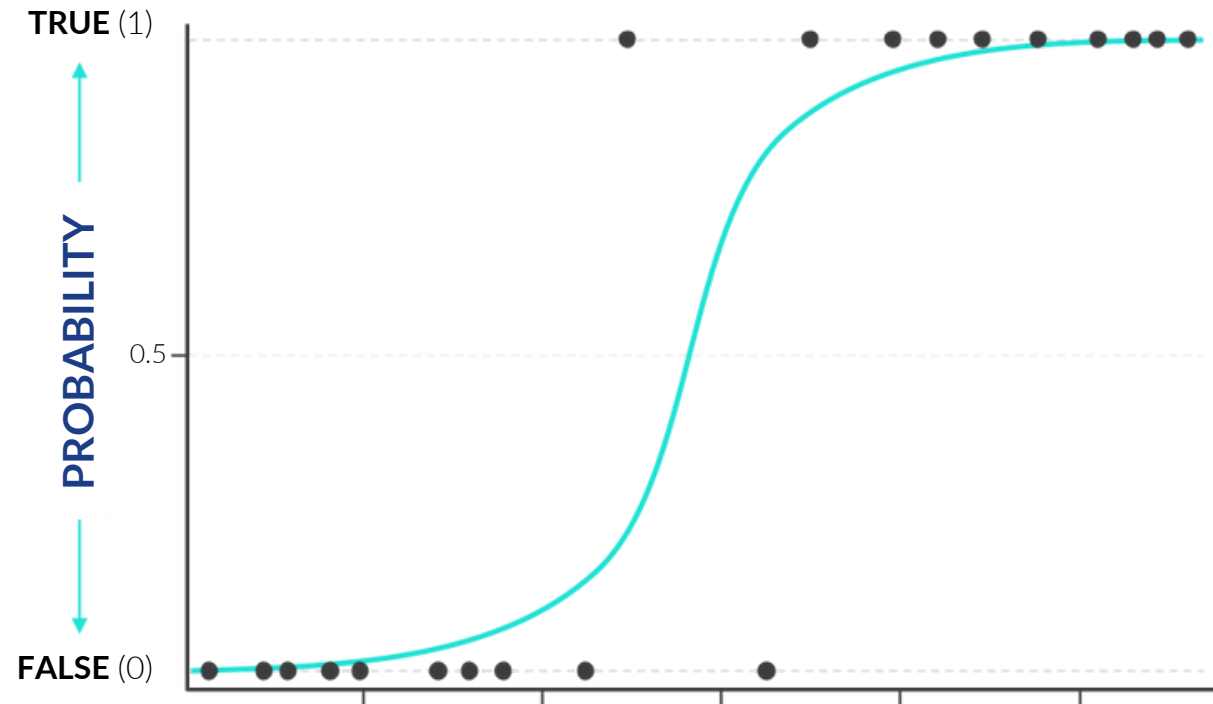


# LOGISTICS REGRESSION

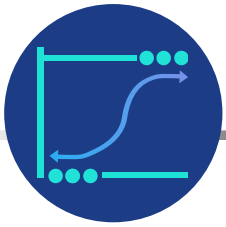
Logistics Regression

Goals of Regression

Evaluation



- Logistic regression plots the **best-fitting curve between 0 and 1**, which tells us the probability of Y being TRUE for any given value of X1



# LOGISTICS REGRESSION

Logistics Regression

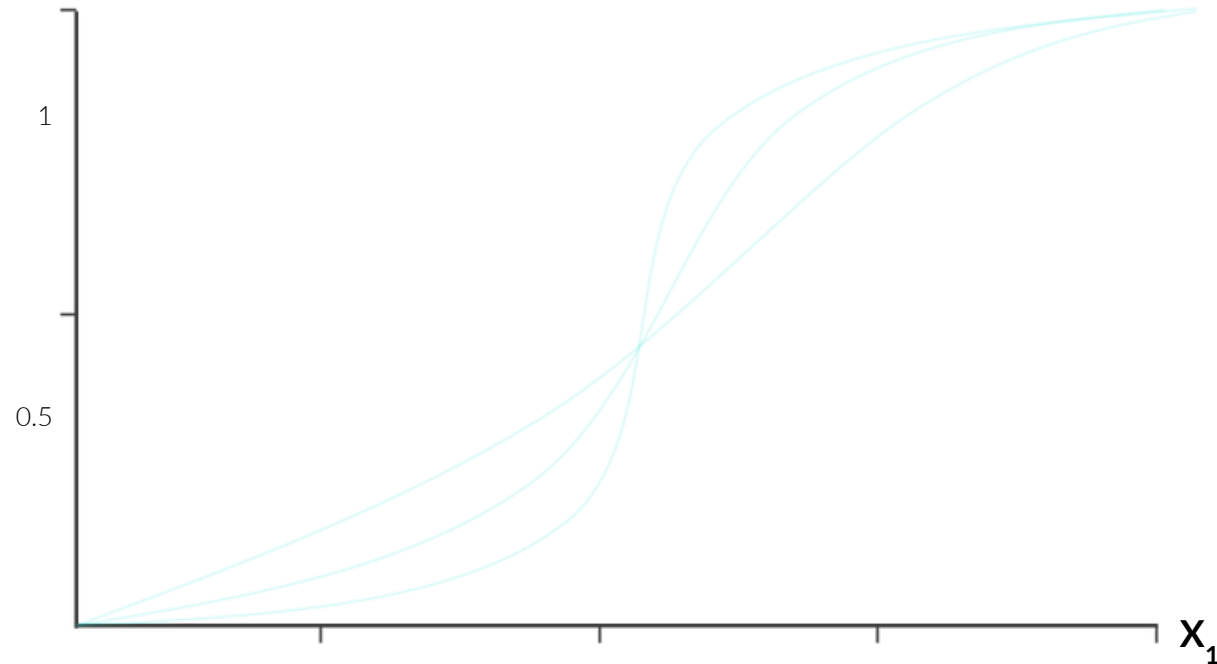
Goals of Regression

Evaluation

Makes the output fall  
between **0** and **1**

**1**

$$\frac{1}{1 + e^{-(\beta_0 + \beta_1 x_1)}}$$

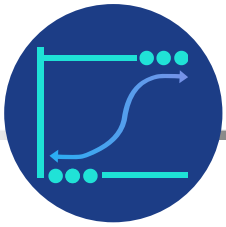


$\beta_1 = 0.5$

$\beta_1 = 1.0$

$\beta_1 = 5.0$

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# LOGISTICS REGRESSION

Logistics Regression

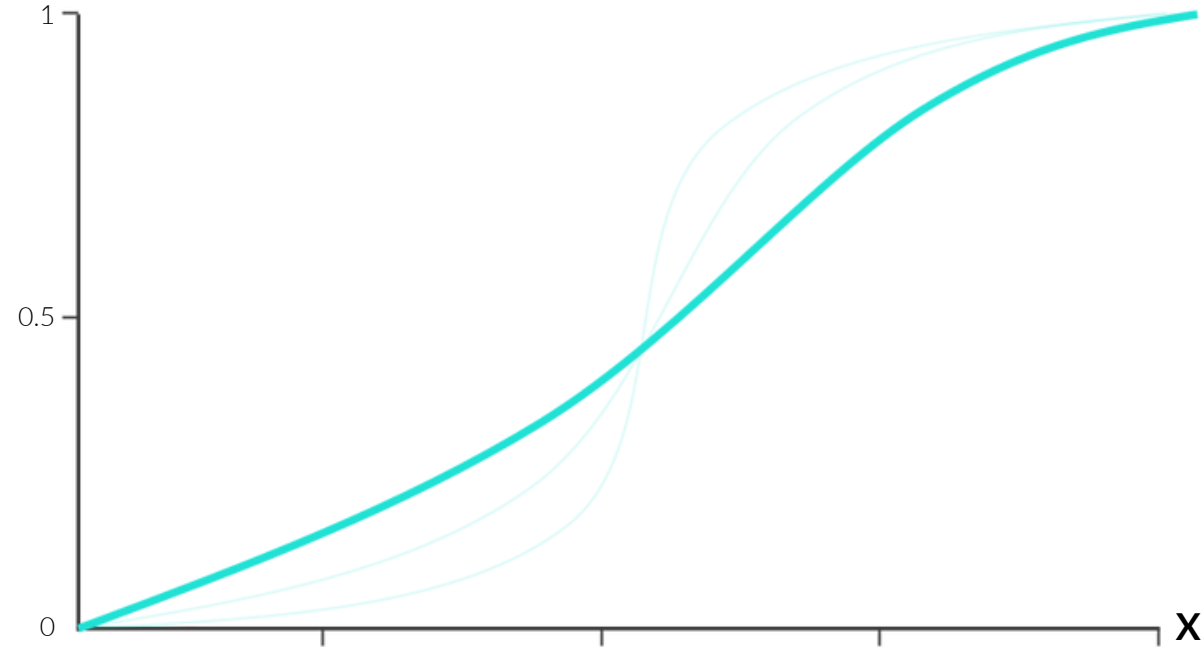
Goals of Regression

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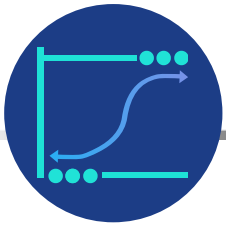
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# LOGISTICS REGRESSION

Logistics Regression

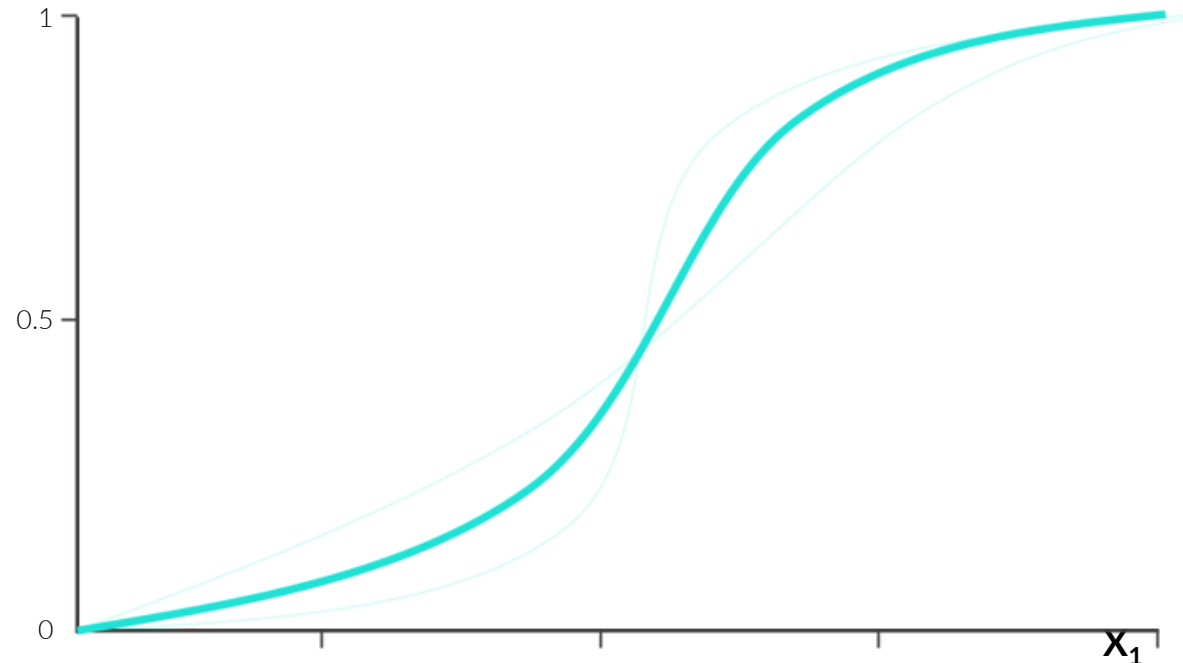
Goals of Regression

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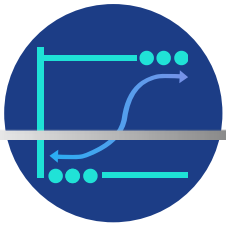


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# GOALS OF REGRESSION

Regression models are used for two primary goals: **prediction** and **inference**

The goal shapes the modeling approach, including the regression algorithm used, the complexity of the model, and more

Logistics Regression

Goals of Regression

Evaluation



## PREDICTION

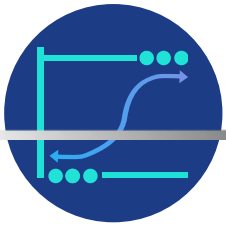
Used to **predict** the target as accurately as possible

*“What is the predict charges for a client given their age?”*



## INFERENCE

- Used to **understand the relationships** between the features and target
- *“How much do a age impact its charges?”*



# EVALUATION

Logistics Regression

Goals of Regression

Evaluation

		Predicted Classes	
		Negative	Positive
Actual Classes	Negative	True Negative	False Positive
	Positive	False Negative	True Positive