

# Logistic Regression

Example - A customer purchase decision for onion during her visits to a vegetable store are shown below in the table.

Visit Index	1	2	3	4	5	6	7	8	9	10
Price	2	2	2	2	2	3	3	3	3	4
Decision	Y	Y	Y	N	Y	Y	Y	N	N	Y

Visit Index	11	12	13	14	15	16	17	18	19	20
Price	4	4	4	4	4	5	5	5	5	5
Decision	N	N	Y	N	N	N	N	N	N	Y

# Logistic Regression

$$(p_2 \ p_2 \ p_2(1-p_2) \ p_2) (p_3^2 (1-p_3)^2) (p_4^2 (1-p_4)^4)$$

Example - A customer purchase decision for onion during her visits to a vegetable store are shown below in the table.

$$p_5 (1-p_5)^4$$

Visit Index	1	2	3	4	5	6	7	8	9	10
Price	2	2	2	2	2	3	3	3	3	4
Decision	Y	Y	Y	N	Y	Y	Y	N	N	Y

Visit Index	11	12	13	14	15	16	17	18	19	20
Price	4	4	4	4	4	5	5	5	5	5
Decision	N	N	Y	N	N	N	N	N	N	Y

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Price	No. of Visits	No. of Purchases	Probability of Purchase
2	5	4	0.8
3	4	2	0.5
4	6	2	0.33
5	5	1	0.2

**When the price is 6, what is the probability that the person will make a purchase?**

Price	No. of Visits	No. of Purchases	Probability of Purchase
2	5	4	0.8
3	4	2	0.5
4	6	2	0.33
5	5	1	0.2

# Ideas to answer the problem

**Objective** - Find the probability of purchase as a function of price

$$P(\text{purchase}) = f(\text{Price})$$

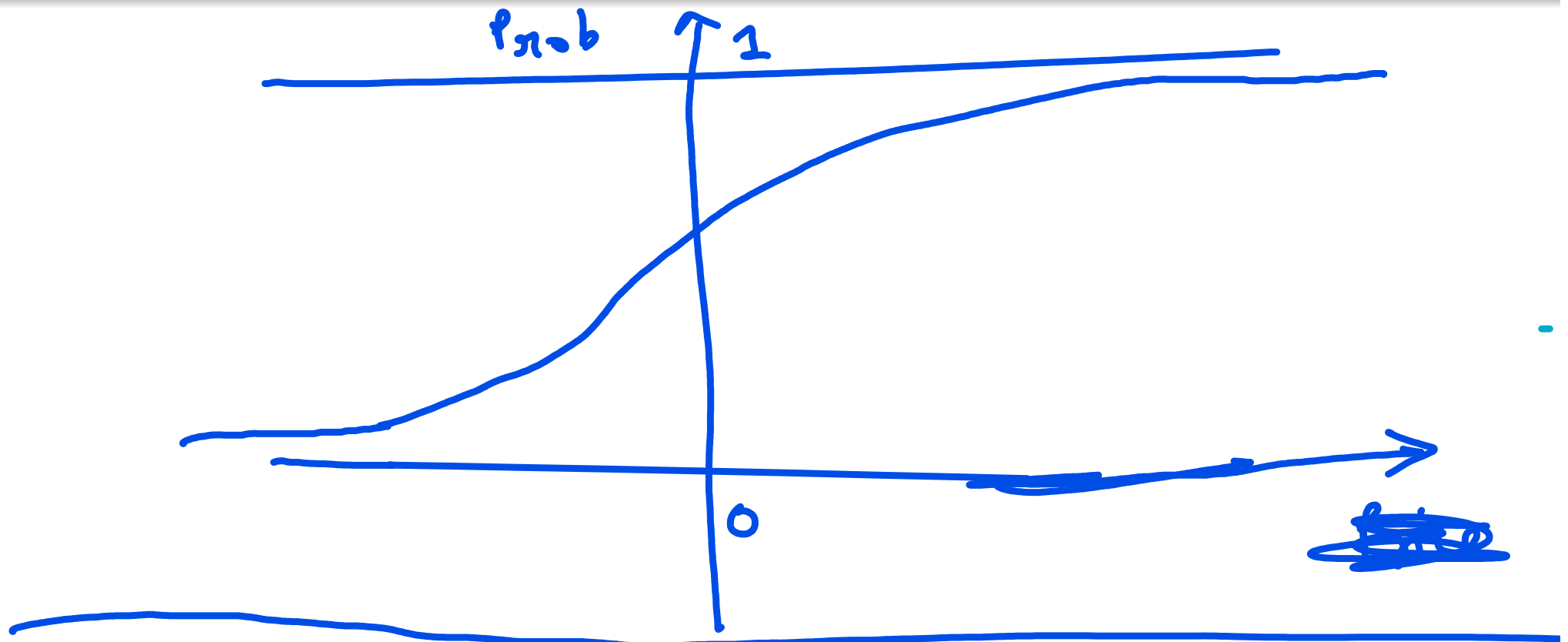
Can we make use of linear regression??

Attempt for a non-linear fit???

# Limitations of Linear Regression

- Doesn't explicitly recognize 0-1 nature of the response
- The impact of change in price on probability of purchase decision is different at different levels of price
- Assumed Linear regression equation treats it as a constant

# Logistic Regression - Main Ideas



$$\frac{e^x}{1 + e^x} = y$$

$$\frac{dy}{dx} = y(1-y)$$

$$\frac{e^{-x}}{(1 + e^{-x})} \left[ \frac{1}{(1 + e^{-x})} \right]$$



# Maximum Likelihood Principle - Ideas

Rain Prediction Model				
	Monday	Tuesday	Wednesday	Thursday
Model A	0.3	0.6	0.7	0.2
Model B	0.2	0.4	0.9	0.6
Rained?	N	Y	Y	N

Model C      0.1      0.6      0.6      0.3

Which Model is better??

Model D      0.1      0.9      0.9      0.1

# Maximum Likelihood Principle - Example

Toss a coin 100 times

You get 70 Meab.

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What "prob" of Heads will maximize the chances of observing what you observed ??

$$p_n^{20} (1-p_n)^{30}$$

Math:

$$70 p n^{19} (1-pn)^{30} + (-30) p n^{20} (1-pn)^{29} = 0$$

# Logistic Regression - Implementation in Python

## Multinomial Regression

# Multinomial Regression - Softmax Function

# Softmax Function - Implementation in Python