

* Deal with Zero Probability

- if $P(x_i/y) = 0$ for some $x_i = x_i$
 & $y = y_i$ it will make posterior 0,
 → To avoid this, we apply "smoothing"

$$P(x_i/y) = \frac{\# x_i \cap y + \alpha}{\# y + \alpha \cdot k}$$

small constant

↑ No. of examples

$P(x_i/y) \geq 0$

* Variables are continuous

	M	M	H	M	M
x_1 (Temp)	30.5	32.7	35.8	30.5	33.4
x_2 (Humidity)	80%	70%	80%	65%	75%
y (Yes/No)	Yes	No	Yes	Yes	No

This does not work

$$P(x_1/y) : P(x_1=30.5/y=Yes)$$

$$P(x_1=32.7/y=Yes)$$

$$P(x_1=35.8/y=Yes)$$

⋮

Idea 1

Divide variable values into different categories

e.g temp = High if $t > 35^{\circ}$
= Med if $30^{\circ} < t < 35^{\circ}$
= Low if $t < 30^{\circ}$

Multinomial Naive Bayes

Variables are multinomial / (binary)
can take a few finite labels.

→ Model parameters are

$$P(x_i / y), \quad P(y)$$

Gaussian Naive Bayes

→ Variables are continuous. e.g.
temp value, pressure value

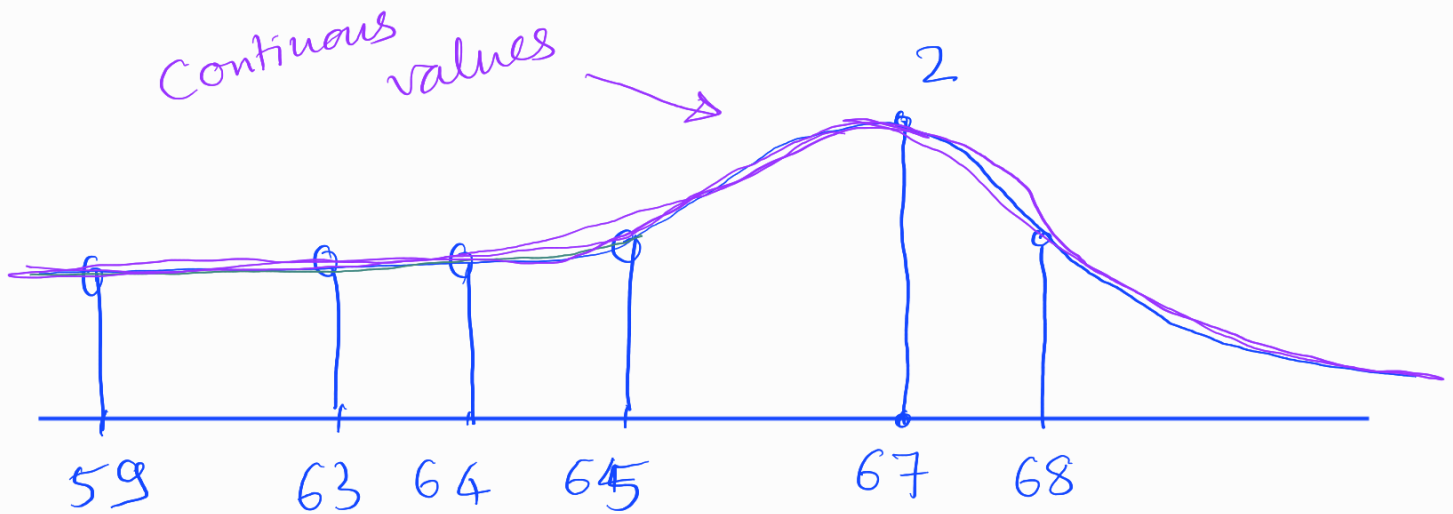
→ Counting probability for all values
of a R.V. is not possible

X_1 (Temp)	30.5	32.7	35.8	30.5	33.4
X_2 (Humidity)	80%	70%	80%	65%	75%

✓ (Yes/No) Yes No Yes Yes No

- X_1 & X_2 are Random Variables with continuous values.
- $P(X_1/y)$ & $P(X_2/y)$ are coming from an underlying Normal Distribution.
- Distribution $P(X_i/y)$ is Normal

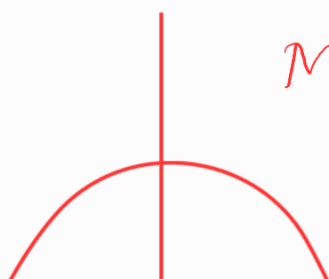
R 65" J 67" N 68" G 59" M 63"
K = 64" A 67



For Normal Distribution the Graph is defined by μ & σ

Bell curve

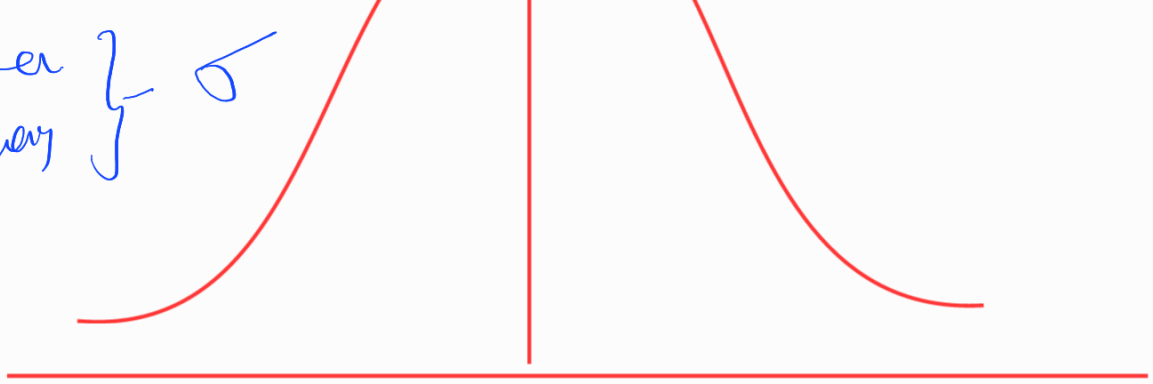
Center = μ



Mean

Standard Deviation

Diameter
1 penny } σ



μ - Mean

Standard Deviation

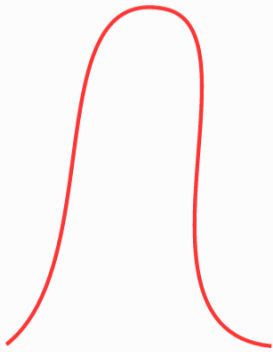
$$\mu = \frac{\sum_{i=1}^n x_i}{N}$$

\rightarrow Average value of x_i

$$S.D. = \sqrt{\frac{\sum (x_i - \mu)^2}{N}}$$

How far are the values from the average

Low SD



High SD



Gaussian Naive Bayes

Parameters

$\mu_{hp} = 62,$

$p(x_i/y) \Rightarrow \mu_{x_iy}, \sigma_{x_iy}$

$\mu_{hp} = 66.7$

h	59	63	64	65	67	68
g	F	F	F	M	M	M

$p(h/g) \leftarrow$ Gaussian

g class label: F/M

h is continuous

Parameters

$$P(h/g=F) = \mu_{hF}, \sigma_{hF}$$

$$P(h/g=M) = \mu_{hM}, \sigma_{hM}$$

