

Session - 5.4

• Feature Selection -

- feature \rightarrow i/p column
- target \rightarrow o/p "

• we don't use all the features rather only use some. this is feature selection

• Why to do feature selection -

1) Curse of dimensionality -

↳ In model only a certain number of feature will give optimum result:

If we increase # features then due to sparsity result good result will not come

2) Computational complexity

↳ reduce column reduce space & time complexity

3) Interpretability -

↳ if less columns then we can interpret easily

• Techniques/types of F.S

↳ filter based technique

↳ wrapper method

↳ Embedded technique

↳ hybrid technique

1) Filter based technique -

↳ we do some statistical measure in each column individually and filter the columns.

In this we have -

- Variance Threshold
- Correlation
- Anova
- Chi square
- mutual info

Step 1

	f_1	f_2	f_3	o/b
1	1	2	1	
2	2	6	2	
3	3	3	3	

→ Here we see that f_1 & f_3 are duplicates
So we remove them

Step 2

1. Variance Threshold

↳ It is applied on two types of features

1) Constant

A	B	C	
2	1	X	Here we see that B column is constant so it has no variance hence remove it.
7	1	N	
2	1	N	

2) Quasi constant feature

→ assume a feature in which ~~is~~ out of 1000 rows in 995 rows value is 1 and in remaining 5 rows value is 0. so variance is 0 so remove it.

- Step →
- decide threshold (like 0.1)
 - for all feature calculate threshold variance
 - drop if var. is less than threshold

→ There are some points to be considered -

- we only focus on individual column. Suppose f_1 has high high var. So we kept it but what if it doesn't have any relatⁿ with o/p and vice-versa means f_2 has low var. So we remove it but it has high relatⁿ with o/p
- we don't see about relatⁿ between f_1 & f_2 what if f_1 has low var. but is related to f_2
- must do scaling
- challenging to find threshold.

→ Correlation (Pearson corr coeff)

↳ we find correlatⁿ b/w i/p & o/p

↳ it is b/w -1 to 1

\uparrow \downarrow
 strong inverse strong positive
 linear relatⁿ linear relatⁿ

→ 0 means no linear relatⁿ (may be non-linear)

↳ we can do -

$f_1 \rightarrow y$
 $f_2 \rightarrow y$
 \vdots
 $-y$

find corr b/w f & y
 if outside this
 (0-0.3 to 0.5) then
 remove feat

or we can find corr b/w f_1, f_2 etc and if having more correlⁿ then keep one of them

Disadvantages -

- It's possible that there is non-linear relatⁿ but we still drop it
- It doesn't capture complex relationship - we only see relatⁿ b/w f_1, f_2 but what if f_1, f_2, f_3 has relatⁿ
- How to know threshold?
- Sensitive to outlier.

→ ANOVA - (Hypothesis test)

i/p	o/p
numerical	categorical (> 2 class)

$f \rightarrow$ o/p (find relatⁿ b/w them)

if strong then keep it.

- strength of relatⁿ is found using f -statics

Disadvantages -

- All groups that are formed should be normally distributed
- Variance of groups must be homogeneous
- Effect of outlier
- Doesn't take account in features selectⁿ

5) Chi square test

↳ Already done

Advantage of Filtered based feature selectⁿ

- Simplicity
- Speed
- Scalability
- pre-processing steps

Disadvantage.

- don't consider interactⁿ b/w features
- model agnostic
- Threshold determination