

Date: \_\_\_\_\_

Muhammad Anas

Day: \_\_\_\_\_

Registration No. :

Course :

Section :

## Question # 01:-

S1: "data science is one of the most important courses in computer science"

S2: "this is one of the best data science courses"

S3: "the data scientists perform data analysis."

Vocabulary: {data, analysis, best, computer, courses, data, important,

BoW: in, is, most, of, one, perform science, scientist, the, this}

	1	2	3
analysis	0	0	1
best	0	1	0
computer	1	0	0
courses	1	1	0
data	1	1	2
important	1	0	0
in	1	0	0
is	1	1	0
most	1	0	0
of	1	1	0
one	1	1	0
perform	0	0	1
science	2	1	0

Date: \_\_\_\_\_

Day: \_\_\_\_\_

scientists	0	0	1
the	1	1	1
this	0	1	0

TF:-

$TF = \frac{\text{term occur in document}}{\text{total terms in document}}$

	analysis	best	computer	courses
1	0.0000	0.0000	0.0833	0.0833
2	0.0000	0.1111	0.0000	0.1111
3	0.1666	0.0000	0.0000	0.0000

	data	important	in	is
1	0.0833	0.0833	0.0833	0.8888
2	0.1111	0.0000	0.0000	0.1111
3	0.3333	0.0000	0.0000	0.0000

	most	of	one	perform
1	0.0833	0.0833	0.0833	0.0000
2	0.0000	0.1111	0.1111	0.0000
3	0.0000	0.0000	0.0000	0.1666

	science	scientists	the	this
1	0.1666	0.0000	0.0833	0.0000
2	0.1111	0.0000	0.1111	0.1111
3	0.0000	0.1666	0.1666	0.0000



## IDF:-

$$IDF = \log \left( \frac{\text{total number of documents}}{\text{number of document have term}} \right)$$

analysis	1.4771
best	1.4771
computer	1.4771
courses	1.1760
data	1.0000
important	1.4771
in	1.4771
is	1.1760
most	1.4771
of	1.1760
one	1.1760
perform	1.4771
science	1.1760
scientists	1.4771
the	1.0000
this	1.4771

## TF-IDF:

$$TF-IDF = (T.F) \times (IDF)$$

	analysis	best	computer	courses
1	0.0000	0.0000	0.3274	0.2490
2	0.0000	0.4229	0.0000	0.3216
3	0.4591	0.0000	0.0000	0.0000

Date: \_\_\_\_\_

Day: \_\_\_\_\_

	data	important	isn	is
1	0.1934	0.3247	0.3274	0.2490
2	0.2498	0.0000	0.0000	0.3216
3	0.5423	0.0000	0.0000	0.0000

	most	of	one	perform
1	0.3274	0.2490	0.2490	0.0000
2	0.0000	0.3216	0.3216	0.0000
3	0.0000	0.0000	0.0000	0.4591

	science	scientists	the	this
1	0.4981	0.0000	0.1934	0.0000
2	0.3216	0.0000	0.2498	0.4229
3	0.0000	0.4591	0.2711	0.0000

## Question # 02 :-

Similarity using Cosine:

$$\text{Cosine}(S_1, S_2) = \frac{\text{dotproduct}(TF \cdot IDF(S_1) \cdot (TF \cdot IDF(S_2)))}{|S_1| \cdot |S_2|}$$

	1	2	3
1	1.0000	0.7126	0.2834
2	0.7126	1.0000	0.3535
3	0.2834	0.3535	1.0000



Similarity using Manhattan:

$$\text{Manhattan}(S_1, S_2) = \sum (\text{abs}(\text{abs}(\text{TF} \cdot \text{IDF}(S_1)) \cdot (\text{TF} \cdot \text{IDF}(S_2)))$$

$$\text{Manhattan}(S_1, S_2) = 0.241223$$

Similarity using Euclidean:-

$$\text{Euclidean}(S_1, S_2) = \sqrt{\sum (\text{TF} \cdot \text{IDF}(S_1) - \text{TF} \cdot \text{IDF}(S_2))^2}$$

$$\text{Euclidean}(S_1, S_2) = 1.362384$$