Assignment No.2

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Div:-CS2

Roll NO.:-08

Sub:-EDS

Dataset: ICC t20 world cup

dataset

Match No.	-1	1st Team	2nd Team				sic First II	-				Method Won by	Winnii	ng Ma Top Score				
1st	Saturday, Jui Grand Prai		United Stat	P	United Stat			194	5	197		Normal Mat Wickets		7 Aaron Jor		94 Dilon Hey		1
2nd	Sunday,June Providence				Afghanistar			95	10	101		Normal Mat Wickets		7 Gulbadin		49 Fazalhaq		3
3rd	Sunday,Jun€ Kensingtor		Namibia	Group B	Namibia	Fielding		109	10	109		Normal Mat Wickets		4 Jan Frylin		45 Mehran K		3
4th	Monday,Jur Nassau Co	u Sri Lanka	South Africa	Group D	Sri Lanka	Batting		77	10	80		Normal Mat Wickets		6 Quinton o		20 Anrich No	ortj	4
5th	Monday, Jun Providence	Afghanista	n Uganda	Group C	Uganda	Fielding		183	5	58		Normal Mat Runs		125 Rahmanu	lla	76 Fazalhaq		5
6th	Tuesday,Jun Kensingtor	Scotland	England	Group B	Scotland	Batting		90	0 Rain	Rain		Rain Rain	Rain	Rain	Rain	Rain	Rain	Rain
7th	Tuesday, Jun Grand Prai	ri Nepal	Netherland	Group D	Netherland	Fielding		106	10	109	4 Netherlands	Normal Mat Wickets		6 Max O'Do	w	54 Tim Pring	le	3
8th	Wednesday Nassau Co	u Ireland	India	Group A	India	Fielding		96	10	97	2 India	Normal Mat Wickets	8 Rohit Sharm		52 Jasprit Bu	ım	2	
9th	Wednesday Providence	Papua New	Uganda	Group C	Uganda	Fielding		77	10	78	8 Uganda	Normal Mat Wickets		3 Riazat Ali	Sh	33 Alpesh Ra	am _.	17
10th	Wednesday Kensington	Australia	Oman	Group B	Oman	Fielding		164	5	125	9 Australia	Normal Mat Runs	39 Marcus Stoi		67 Marcus St	toi	3	
11th	Thursday,Ju Grand Prai	ri Pakistan	United Stat	Group A	United Stat	Fielding		159	7	160	3 United State	Normal Mat Runs		1 Monank F	at	50 Nosthush	Kε	30
12th	Thursday,Ju Kensingtor	Namibia	Scotland	Group B	Namibia	Batting		155	9	157	5 Scotland	Normal Mat Wickets		5 Michael L	ea	35 Brad Curr	ie	2
13th	Friday,June Nassau Co	u Canada	Ireland	Group A	Ireland	Fielding		137	7	125	7 Canada	Normal Mat Runs	12 Nicholas Kir		49 Jeremy G	ore	2	
14th	Friday,June Providence	Afghanista	n New Zealar	Group C	New Zealar	Fielding		159	6	75	10 Afghanistan	Normal Mat Runs		84 Rahmanu	lla	80 Rashid Kh	nar	4
15th	Friday,June Grand Prai	ri Sri Lanka	Bangladesh	Group D	Bangladesh	Fielding		124	9	125	8 Bangladesh	Normal Mat Wickets		2 Pathum N	list	47 Rishad Ho	oss	3
16th	Saturday, Jui Nassau Co	u Netherland	k South Africa	Group D	South Africa	Fielding		103	9	106	6 South Africa	Normal Mat Wickets		4 David Mil	ler	59 Ottneil Ba	nae	4
17th	Saturday, Jui Kensington	Australia	England	Group B	England	Fielding		201	7	165	6 Australia	Normal Mat Runs		36 Jos Buttle	r	42 Adam Zar	mp	2
18th	Saturday, Jui Providence	West Indie	s Uganda	Group C	West Indies	Batting		173	5	39	10 West Indies	Normal Mat Runs	134 Johnso		Chi	44 Akeal Hos	sei	5
19th	Sunday,June Nassau Co	u India	Pakistan	Group A	Pakistan	Fielding		119	10	113	7 India	Normal Mat Runs	6 Rishabh Par		42 Jasprit Bu	ımı	3	
20th	Sunday,June Sir Vivian F	i Oman	Scotland	Group B	Oman	Batting		150	7	153	3 Scotland	Normal Mat Wickets	7 Brandon Mc		31 Chris Greave		1	
21th	Monday,Jur Nassau Co	u South Afric	a Bangladesh	Group D	South Africa	Batting		113	6	109	7 South Africa	Normal Mat Runs		4 Heinrich H	(la	46 Tanzim H	asa	3
22th	Tuesday, Jun Nassau Co	u Canada	Pakistan	Group A	Pakistan	Fielding		106	7	107	3 Pakistan	Normal Mat Wickets		7 Aaron Joh	nns	52 Mohamm	nac	2
23th	Tuesday,Jur Central Bro	Nepal	Sri Lanka	Group D	Rain	Rain	Rain	Rain	Rain	Rain	Rain	Rain Rain	Rain	Rain	Rain	Rain	Rain	Rain
24th	Tuesday,Jur Sir Vivian F	i Namibia	Australia	Group B	Australia	Fielding		72	10	74	1 Australia	Normal Mat Wickets		9 Gerhard E	ra	36 Adam Zar	mp	4
25th	Wednesday Nassau Co	u United Stat	te India	Group A	India	Fielding		110	8	111	3 India	Normal Mat Wickets		7 Suryakum	ar	50 Arshdeep	Si	4
26th	Wednesday Brian Lara	S West Indie:	s New Zealar	Group C	New Zealar	Fielding		149	9	136	9 West Indies	Normal Mat Runs		13 Sherfane	Ru	68 Trent Bou	ult	3
27th	Thursday,Ju Arnos Vale	(Bangladesh	Netherland	Group D	Netherland	Fielding		159	5	134	8 Bangladesh	Normal Mat Runs		25 Shakib Al	На	64 Paul van I	Μe	2
28th	Thursday,Ju Sir Vivian F	i Oman	England	Group B	England	Fielding		47	10	50	2 England	Normal Mat Wickets		8 Jos Buttle	r	24 Adil Rashi	id	4
29th	Thursday,Ju Brian Lara	S Papua New	Afghanistar	Group C	Afghanistar	Fielding		95	10	101	3 Afghanistan	Normal Mat Wickets		7 Gulbadin	Na	49 Fazalhag	Fa	3
30th	Friday,June Central Bri	United Stat	te Ireland	Group A	Rain	Rain	Rain	Rain	Rain	Rain	Rain	Rain Rain	Rain	Rain	Rain	Rain	Rain	Rai
31th	Friday,June Arnos Vale	(South Afric	a Nepal	Group D	Nepal	Fielding		115	7	114	7 South Africa	Normal Mat Runs		1 Reeza He	nd	43 Tabraiz Sl	hai	4
32th	Friday,June Brian Lara	S Uganda	New Zealar	Group C	New Zealar	Fielding		40	10	41	1 New Zealan	Normal Mat Wickets		9 Devon Co	nv	22 Tim South	ner	3

```
[ ] import pandas as pd
import numpy as np
df = pd.read_csv('/icc t20 world cup.csv')
            def unique_venues(df):
    return df['Venue'].unique()
 Unique Venues: ['Grand Prairie Stadium, Dallas' 'Providence Stadium, Guyana'
'Kensington Oval, Bridgetown, Barbados'
'Nassau County International Cricket Stadium, New York'
'Sir Vivian Richards Stadium, North Sound, Antigua'
'Central Broward Regional Park Stadium Turf Ground, Lauderhill'
'Brian Lara Stadium, Tarouba, Trinidad'
'Arnos Vale Ground, Kingstown, St Vincent'
'Daren Sammy National Cricket Stadium, Gros Islet, St Lucia']
  # Problem 2: Count of matches played in each stage
def stage_match_count(df):
    return df['Stage'].value_counts()
  # Problem 2: Count of matches played in each stage
def stage_match_count(df):
    return df['Stage'].value_counts()
 [ ] # Problem 3: Team with most match wins def most_match_wins(df):
    return df['winners'].value_counts().idxmax()
  Win Type Count:
Won by

→ Runs 26
Wickets 25
Rain 4
Name: count, dtype: int64
                                                                                                                                                                                                                                                                                                                                                  ↑ ↓ + © ‡ 🖟 🖽 :
  # Problem 5: Toss wins by each team
def toss_winners_count(df):
    return df['Toss Winning'].value_counts()
```

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↑ ↓ + ⊕ ‡ ᡚ Ⅲ :
Problem 7: Count of normal matches def normal_matches_count(df):
    return df[df['Method'] == 'Normal Match'].shape[0]
 Two Number of Normal Matches: 48
[ ] # Problem 8: Player of the match awards count

def player_of_match_counts(df):

return df['Player Of The Match'].value_counts()
Player of the Match Counts:
Player Of The Match
Rain

Fazalhaq Farcoci

Jasprit Bumrah

2

Marcus Stoinis

2

Quinton de Kock

2

Tabraiz Shamsi

2

Adil Rashid

2

Anrich Nortje

David Miese

1
10p 5 Best Bowlers:
Best Bowler
Rain
Fazalhag Farooqi 3
Rashid Khan 3
Adil Rashid 3
Jasprit Bumrah 2
Name: count, dtype: int64
# Problem 10: Variance of Second Innings Scores
def var.second.innings(df):
scores = pd.to_numeric(df['Second Innings Score'], errors='coerce')
return np.var(scores)
[ ] # Problem 11: All unique toss decisions def unique_toss_decisions(df): return df['Toss Decision'].unique()
 ### a Problem 12: Most frequent 1st team

def most_frequent_first_team(df):
    return df['1st Team'].value_counts().idxmax()
 → Total Matches Played: 55
[ ] # Problem 13: Most frequent 2nd team

def most_frequent_second_team(df):
    return df['2nd Team'].value_counts().idxmax()
 → Most Frequent 2nd Team: England
[ ] # Problem 14: Number of matches played by India
def india_matches_count(df):
    return df['1st Team'].str.contains('India').sum() + df['2nd Team'].str.contains('India').sum()
  Tr Number of Matches Played by India: 9

■ Problem 15: Matches using D/L method
def dl_method_matches(df):
    return df[df['Method'] == 'D/L Method'].shape[0]

→ Number of Matches Using D/L Method: 0
```

	### Problem 22: List all unique values of Winning Margin def unique winning margins(df): ###################################
	Unique Winning Margins: [1. 2. 3. 4. 5. 6. 7. 8. 9. 10. 12. 13. 18. 21. 24. 25. 28. 36. 39. 41. 47. 50. 68. 83. 84. 104. 125. 134.]
	# Problem 23: Calculate the median of First Innings Scores def median_first_innings(df): scores = qb.to_numeric(df)'rirst Innings Score'], errors='coerce').dropna().values return np.median(scores) print("Median First Innings Score:", median_first_innings(df))
	Median First Innings Score: 131.5
	# Problem 24: Check if there are any duplicate First Innings Scores def has duplicate scores(df): scores = qh to _numeric(df)' First Innings Score'), errors='coerce').dropna().values return len(scores) != len(np.unique(scores)) print("Has Duplicate Scores:", has_duplicate_scores(df))
⊋	Has Duplicate Scores: True
0	# Problem 25: Count how many First Innings Scores are between 100 and 160 def count_scores_in_ramge(df): scores = pd.to_numeric(df)'First Innings Score'], errors='coerce').dropna().values return np.sum((scores >= 160) & (scores <= 160)) print("Scores between 100 and 160:", count_scores_in_range(df))
	Scores between 100 and 160: 26
	# Problem 26: Find the index of the match with the highest Winning Margin def index_max_margin(df): margins = pd.to_numeric(df['Winning Margin'], errors='coerce').dropna().values return np.argmax(margins)
	print("Index of Highest Winning Margin:", index_max_margin(df)) Index of Highest Winning Margin: 16
	# Problem 27: Create a boolean array showing if each Winning Margin is greater than 5 def margin above_5_boolean(df): margins = pd.to_numeric(df['Winning Margin'], errors='coerce').dropna().values return margins > 5: print("Margins > 5:", margin_above_5_boolean(df))
₹	Margins > 5: [True True False True True True True False True False False True True False False True True True True True True True Tru
(0)	# Problem 28: Normalize the Winning Margins using Min-Max normalization def normalize_margins(df):
	on no mailze_mengins(u); margins = pdt.o numeric(df['Winning Margin'], errors='coerce').dropna().values return (margins - np.min(margins)) / (np.max(margins) - np.min(margins)) print('Normalized Margins:', normalize_margins(df))
æ	Normalized Margins: [0.40511278 0.60511278 0.60255639 0.63759398 0.93233083 0.63759398 0.05253138 0.0563135 0.0563739 0.28571429 0. 0. 0.2080015 0.60276677 0.05253138 0.05253799 0.05511278 0.05255639 0.6511278 0.05255639 0.05511278 0.0601598 0.05511278 0.0601598 0.05511278 0.0601598 0.05511278 0.0601598 0.05511278 0.0601598 0.05511278 0.0601598 0.05511278 0.0601598 0.05511278 0.0601598 0.05511278 0.0601598 0.05511278 0.05601598 0.05511278 0.05601598 0.056015
	# Problem 29: Calculate average First Innings Score for matches with Winning Margin > 50 def avg score high margin(df): scores = pd.to numeric(df' First Innings Score'], errors='coerce').dropna().values margins = pd.to numeric(df' Winning Margin'], errors='coerce').dropna().values min_len = min(len(scores), len(margins)) return np.mean(scores!:min_len][margins!:min_len] > 50]) print('Average Score for High Murgin Mutches:', avg_score_high_margin(df))
	Average Score for High Margin Matches: 126.833333333333
	<pre># Problem 30: Sort the First Innings Scores in ascending order def sort_first_imnings_scores(df): scores = pd.to_numeric(df'First_Innings_Score'], errors='coerce').dropna().values return np.sort(scores) print("sorted First_Innings_Scores:", sort_first_innings_scores(df))</pre>
Ð	Sorted First Innings Scores: [40. 47. 56. 72. 77. 77. 78. 90. 95. 95. 96. 103. 106. 106.