

#READING DATA

```
import pandas as pd
matches = pd.read_csv("matches.csv", index_col=0)
matches.head()
```

	date	time	comp	round	day	venue	result	gf	ga	opponent	...	match report
1	2021-08-15	16:30	Premier League	Matchweek 1	Sun	Away	L	0.0	1.0	Tottenham	...	Match Report
2	2021-08-21	15:00	Premier League	Matchweek 2	Sat	Home	W	5.0	0.0	Norwich City	...	Match Report
3	2021-08-28	12:30	Premier League	Matchweek 3	Sat	Home	W	5.0	0.0	Arsenal	...	Match Report
4	2021-09-04	15:00	Premier League	Matchweek 4	Sun	Away	L	0.0	1.0	Leicester	...	Match Report

```
matches.shape
```

```
(1389, 27)
```

#INVESTGATING MISSING DATA

```
#Should have 1520 matches in 2 seasons of EPL but here 1389 match data are present so let's figure out what's going on here
#as we know 3 teams get relegated in EPL each year so 6 teams should have less amount of matches.
```

```
matches["team"].value_counts()
```

```
Southampton          72
Brighton and Hove Albion 72
Manchester United      72
West Ham United        72
Newcastle United       72
Burnley                71
Leeds United           71
Crystal Palace         71
Manchester City         71
Wolverhampton Wanderers 71
Tottenham Hotspur      71
Arsenal                71
Leicester City         70
Chelsea                70
Aston Villa            70
Everton                70
Liverpool              38
Fulham                 38
West Bromwich Albion    38
Sheffield United       38
Brentford              34
Watford                 33
Norwich City           33
Name: team, dtype: int64
```

```
#Liverpool didn't get relegated in recent years so let's see why this team has less amount of matches
```

```
matches[matches["team"] == "Liverpool"]
```

	date	time	comp	round	day	venue	result	gf	ga	opponent	...	match report	notes	sh	sot	dist	fk	pk	pkatt	seas
1	2020-09-12	17:30	Premier League	Matchweek 1	Sat	Home	W	4.0	3.0	Leeds United	...	Match Report	NaN	20.0	4.0	17.0	0.0	2.0	2.0	20
2	2020-09-20	16:30	Premier League	Matchweek 2	Sun	Away	W	2.0	0.0	Chelsea	...	Match Report	NaN	17.0	5.0	17.7	1.0	0.0	0.0	20
4	2020-09-28	20:00	Premier League	Matchweek 3	Mon	Home	W	3.0	1.0	Arsenal	...	Match Report	NaN	21.0	9.0	16.8	0.0	0.0	0.0	20
6	2020-10-04	19:15	Premier League	Matchweek 4	Sun	Away	L	2.0	7.0	Aston Villa	...	Match Report	NaN	14.0	8.0	15.8	1.0	0.0	0.0	20
7	2020-10-17	12:30	Premier League	Matchweek 5	Sat	Away	D	2.0	2.0	Everton	...	Match Report	NaN	22.0	8.0	15.0	1.0	0.0	0.0	20
9	2020-10-24	20:00	Premier League	Matchweek 6	Sat	Home	W	2.0	1.0	Sheffield Utd	...	Match Report	NaN	17.0	5.0	18.2	1.0	0.0	0.0	20
11	2020-10-31	17:30	Premier League	Matchweek 7	Sat	Home	W	2.0	1.0	West Ham	...	Match Report	NaN	8.0	2.0	18.6	1.0	1.0	1.0	20
13	2020-11-08	16:30	Premier League	Matchweek 8	Sun	Away	D	1.0	1.0	Manchester City	...	Match Report	NaN	9.0	2.0	21.5	0.0	1.0	1.0	20
14	2020-11-22	19:15	Premier League	Matchweek 9	Sun	Home	W	3.0	0.0	Leicester City	...	Match Report	NaN	24.0	12.0	11.9	0.0	0.0	0.0	20
16	2020-11-28	12:30	Premier League	Matchweek 10	Sat	Away	D	1.0	1.0	Brighton	...	Match Report	NaN	6.0	2.0	20.9	0.0	0.0	0.0	20
18	2020-12-06	19:15	Premier League	Matchweek 11	Sun	Home	W	4.0	0.0	Wolves	...	Match Report	NaN	11.0	6.0	16.6	1.0	0.0	0.0	20
20	2020-12-13	16:30	Premier League	Matchweek 12	Sun	Away	D	1.0	1.0	Fulham	...	Match Report	NaN	11.0	5.0	20.0	1.0	1.0	1.0	20
21	2020-12-16	20:00	Premier League	Matchweek 13	Wed	Home	W	2.0	1.0	Tottenham	...	Match Report	NaN	17.0	11.0	15.5	0.0	0.0	0.0	20
22	2020-12-19	12:30	Premier League	Matchweek 14	Sat	Away	W	7.0	0.0	Crystal Palace	...	Match Report	NaN	14.0	7.0	13.2	1.0	0.0	0.0	20
23	2020-12-27	16:30	Premier League	Matchweek 15	Sun	Home	D	1.0	1.0	West Brom	...	Match Report	NaN	17.0	2.0	17.8	2.0	0.0	0.0	20
24	2020-12-30	20:00	Premier League	Matchweek 16	Wed	Away	D	0.0	0.0	Newcastle Utd	...	Match Report	NaN	11.0	4.0	16.7	0.0	0.0	0.0	20
25	2021-01-04	20:00	Premier League	Matchweek 17	Mon	Away	L	0.0	1.0	Southampton	...	Match Report	NaN	17.0	1.0	14.3	0.0	0.0	0.0	20
	2021		Premier League	Matchweek 18						Manchester City	...	Match Report	NaN							20

#so we can see that we are missing data of this season (22-23).
matches["round"].value_counts()

Matchweek 1	39
Matchweek 16	39
Matchweek 34	39
Matchweek 32	39
Matchweek 31	39
Matchweek 29	39
Matchweek 28	39
Matchweek 26	39
Matchweek 25	39
Matchweek 24	39
Matchweek 23	39
Matchweek 2	39
Matchweek 19	39
Matchweek 17	39
Matchweek 20	39
Matchweek 15	39
Matchweek 5	39
Matchweek 3	39
Matchweek 13	39
Matchweek 12	39
Matchweek 4	39
Matchweek 11	39
Matchweek 10	39
Matchweek 9	39
Matchweek 8	39
Matchweek 14	39
Matchweek 7	39
Matchweek 6	39
Matchweek 30	37

```

Matchweek 27    37
Matchweek 22    37
Matchweek 21    37
Matchweek 18    37
Matchweek 33    32
Matchweek 35    20
Matchweek 36    20
Matchweek 37    20
Matchweek 38    20
Name: round, dtype: int64

```

#Now we know where our missing rows went, we lack some rows of some matchweeks

#CLEANING DATA FOR MACHING LEARNING

matches.dtypes

```

date           object
time           object
comp           object
round          object
day            object
venue          object
result         object
gf             float64
ga             float64
opponent       object
xg             float64
xga            float64
poss           float64
attendance     float64
captain        object
formation      object
referee        object
match report   object
notes          float64
sh             float64
sot            float64
dist           float64
fk             float64
pk             float64
pkatt          float64
season         int64
team           object
dtype: object

```

#ML algorithms can't work with objects. So we need to convert them to workable functions

```

matches["date"] = pd.to_datetime(matches["date"]) # not creating new column but overwriting the existing one
matches.dtypes

```

```

date           datetime64[ns]
time           object
comp           object
round          object
day            object
venue          object
result         object
gf             float64
ga             float64
opponent       object
xg             float64
xga            float64
poss           float64
attendance     float64
captain        object
formation      object
referee        object
match report   object
notes          float64
sh             float64
sot            float64
dist           float64
fk             float64
pk             float64
pkatt          float64
season         int64
team           object
dtype: object

```

#CREATING PREDICTORS FOR ML

```

matches["venue_code"] = matches["venue"].astype("category").cat.codes #converting strings into categories and converting categories into numb

```

matches

	date	time	comp	round	day	venue	result	gf	ga	opponent	...
1	2021-08-15	16:30	Premier League	Matchweek 1	Sun	Away	L	0.0	1.0	Tottenham	...
2	2021-08-21	15:00	Premier League	Matchweek 2	Sat	Home	W	5.0	0.0	Norwich City	...
3	2021-08-28	12:30	Premier League	Matchweek 3	Sat	Home	W	5.0	0.0	Arsenal	...
4	2021-09-11	15:00	Premier League	Matchweek 4	Sat	Away	W	1.0	0.0	Leicester City	...
6	2021-09-18	15:00	Premier League	Matchweek 5	Sat	Home	D	0.0	0.0	Southampton	...
...
38	2021-05-02	19:15	Premier League	Matchweek 34	Sun	Away	L	0.0	4.0	Tottenham	...
39	2021-05-08	15:00	Premier League	Matchweek 35	Sat	Home	L	0.0	2.0	Crystal Palace	...

```
matches["opp_code"] = matches["opponent"].astype("category").cat.codes
matches
```

	date	time	comp	round	day	venue	result	gf	ga	opponent	...
1	2021-08-15	16:30	Premier League	Matchweek 1	Sun	Away	L	0.0	1.0	Tottenham	...
2	2021-08-21	15:00	Premier League	Matchweek 2	Sat	Home	W	5.0	0.0	Norwich City	...
3	2021-08-28	12:30	Premier League	Matchweek 3	Sat	Home	W	5.0	0.0	Arsenal	...
4	2021-09-11	15:00	Premier League	Matchweek 4	Sat	Away	W	1.0	0.0	Leicester City	...
6	2021-09-18	15:00	Premier League	Matchweek 5	Sat	Home	D	0.0	0.0	Southampton	...
...
38	2021-05-02	19:15	Premier League	Matchweek 34	Sun	Away	L	0.0	4.0	Tottenham	...
39	2021-05-08	15:00	Premier League	Matchweek 35	Sat	Home	L	0.0	2.0	Crystal Palace	...

```
matches["hour"] = matches["time"].str.replace(":.+", "", regex=True).astype("int") #replace the colon and minutes(just keep the hour) with no
matches["day_code"] = matches["date"].dt.dayofweek #changing weekdays with numbers
matches
```

	date	time	comp	round	day	venue	result	gf	ga	opponent	...
1	2021-08-15	16:30	Premier League	Matchweek 1	Sun	Away	L	0.0	1.0	Tottenham	...
2	2021-08-21	15:00	Premier League	Matchweek 2	Sat	Home	W	5.0	0.0	Norwich City	...
3	2021-08-28	12:30	Premier League	Matchweek 3	Sat	Home	W	5.0	0.0	Arsenal	...
4	2021-09-11	15:00	Premier League	Matchweek 4	Sat	Away	W	1.0	0.0	Leicester City	...
6	2021-09-18	15:00	Premier League	Matchweek 5	Sat	Home	D	0.0	0.0	Southampton	...
...
38	2021-05-02	19:15	Premier League	Matchweek 34	Sun	Away	L	0.0	4.0	Tottenham	...
39	2021-05-08	15:00	Premier League	Matchweek 35	Sat	Home	L	0.0	2.0	Crystal Palace	...

```

matches["target"] = (matches["result"] == "W").astype("int") #Set up a target which we are going to predict (e.g. our team won or not)
#win as 1 and draw or loss as 0
matches

```

	date	time	comp	round	day	venue	result	gf	ga	opponent	...
1	2021-08-15	16:30	Premier League	Matchweek 1	Sun	Away	L	0.0	1.0	Tottenham	...
2	2021-08-21	15:00	Premier League	Matchweek 2	Sat	Home	W	5.0	0.0	Norwich City	...
3	2021-08-28	12:30	Premier League	Matchweek 3	Sat	Home	W	5.0	0.0	Arsenal	...
4	2021-09-11	15:00	Premier League	Matchweek 4	Sat	Away	W	1.0	0.0	Leicester City	...
6	2021-09-18	15:00	Premier League	Matchweek 5	Sat	Home	D	0.0	0.0	Southampton	...
...
38	2021-05-02	19:15	Premier League	Matchweek 34	Sun	Away	L	0.0	4.0	Tottenham	...
39	2021-05-08	15:00	Premier League	Matchweek 35	Sat	Home	L	0.0	2.0	Crystal Palace	...

```

#CREATING INITIAL ML MODEL
from sklearn.ensemble import RandomForestClassifier #RandomForest can pick up Non-Linear type of data
rf = RandomForestClassifier(n_estimators=90, min_samples_split=10, random_state=1) #RF-> Series of Decision Trees but each DT has slightly different
#min_samples_split --> number of samples we want to have in a leaf of the DT before splitting the node, the higher this is the less likely to overfit
#RF has lot of random parameters in it. If we set a random state it means if we run the RF multiple times we would get the same result.
train = matches[matches["date"] < '2022-01-01'] #time series data. Make sure All the data in the test set comes after the training set.
test = matches[matches["date"] > '2022-01-01']
#Why split up into train and test?
# We want the algorithm to do well in the predicting future matches which is why we test it out on the data that it hasn't been trained on.
predictors = ["venue_code", "opp_code", "hour", "day_code"] #list of the predictor columns we have created
rf.fit(train[predictors], train["target"]) #fit our RF model. (the .fit method going to train our RF model with predictors)
RandomForestClassifier(min_samples_split=10, n_estimators=50, random_state=1)
preds = rf.predict(test[predictors])
#check accuracy in percentage
from sklearn.metrics import accuracy_score #what percentage of the time was your prediction accurate
acc = accuracy_score(test["target"], preds)
acc

#see in which situation our accuracy was high or low, need to create a Data frame for that
combined = pd.DataFrame(dict(actual=test["target"], prediction=preds))
pd.crosstab(index=combined["actual"], columns=combined["prediction"]) #we can see we were right most of the time about loss or draw but less r

```

	prediction		
	0	1	
actual			
0	141	31	
1	76	28	

```

#Revise our accuracy metric
from sklearn.metrics import precision_score #tells us when we predicted a win, what percentage of time it wins
precision_score(test["target"], preds) #47% is not good, not great precision

```

```
0.4745762711864407
```

```

#IMPROVING PRECISION WITH ROLLING AVERAGES --> using more predictors
#split the matches dataframe up by team. compute the rolling avg. (how many shots per goal, how many threats etc.)
grouped_matches = matches.groupby("team") #creates one dataframe for 1 squad in data.
group = grouped_matches.get_group("Manchester City")
group

```

	date	time	comp	round	day	venue	result	gf	ga	opponent	...
1	2021-08-15	16:30	Premier League	Matchweek 1	Sun	Away	L	0.0	1.0	Tottenham	...
2	2021-08-21	15:00	Premier League	Matchweek 2	Sat	Home	W	5.0	0.0	Norwich City	...
3	2021-08-28	12:30	Premier League	Matchweek 3	Sat	Home	W	5.0	0.0	Arsenal	...
4	2021-09-11	15:00	Premier League	Matchweek 4	Sat	Away	W	1.0	0.0	Leicester City	...
6	2021-09-18	15:00	Premier League	Matchweek 5	Sat	Home	D	0.0	0.0	Southampton	...
...
	2021-		Premier	Matchweek						Crystal	

```
#if we were in matchweek 4 how did City do in previous 3 matches and use that information to feed the algorithm
def rolling_averages(group, cols, new_cols): #new_cols for assigning the rolling avgs to.
    group = group.sort_values("date")
    rolling_stats = group[cols].rolling(3, closed='left').mean() #leaves the row that is going to predicted
    group[new_cols] = rolling_stats
    group = group.dropna(subset=new_cols) #drops missing values
    return group
cols = ["gf", "ga", "sh", "sot", "dist", "fk", "pk", "pkatt"] #cols that is going to be computed rolling avgs for
new_cols = [{"c"}_rolling for c in cols]
new_cols #going to be created rolling avgs in them

['gf_rolling',
'ga_rolling',
'sh_rolling',
'sot_rolling',
'dist_rolling',
'fk_rolling',
'pk_rolling',
'pkatt_rolling']
```

```
rolling_averages(group, cols, new_cols) #Only for Manchester City
```

	date	time	comp	round	day	venue	result	gf	ga	opponent	...	d
5	2020-10-17	17:30	Premier League	Matchweek 5	Sat	Home	W	1.0	0.0	Arsenal	...	
7	2020-10-24	12:30	Premier League	Matchweek 6	Sat	Away	D	1.0	1.0	West Ham	...	
9	2020-10-31	12:30	Premier League	Matchweek 7	Sat	Away	W	1.0	0.0	Sheffield Utd	...	
11	2020-11-08	16:30	Premier League	Matchweek 8	Sun	Home	D	1.0	1.0	Liverpool	...	
12	2020-11-21	17:30	Premier League	Matchweek 9	Sat	Away	L	0.0	2.0	Tottenham	...	
...	
42	2022-03-14	20:00	Premier League	Matchweek 29	Mon	Away	D	0.0	0.0	Crystal Palace	...	
44	2022-04-02	15:00	Premier League	Matchweek 31	Sat	Away	W	2.0	0.0	Burnley	...	

```
#apply this to all of our teams
matches_rolling = matches.groupby("team").apply(lambda x: rolling_averages(x, cols, new_cols))
matches_rolling
```

		date	time	comp	round	day	venue	result	gf	ga	
team											
Arsenal	6	2020-10-04	14:00	Premier League	Matchweek 4	Sun	Home	W	2.0	1.0	
	7	2020-10-17	17:30	Premier League	Matchweek 5	Sat	Away	L	0.0	1.0	N
	9	2020-10-25	19:15	Premier League	Matchweek 6	Sun	Home	L	0.0	1.0	
	11	2020-11-01	16:30	Premier League	Matchweek 7	Sun	Away	W	1.0	0.0	N
	13	2020-11-08	19:15	Premier League	Matchweek 8	Sun	Home	L	0.0	3.0	
...	
Wolverhampton Wanderers	32	2022-03-13	14:00	Premier League	Matchweek 29	Sun	Away	W	1.0	0.0	
	33	2022-03-18	20:00	Premier League	Matchweek 30	Fri	Home	L	2.0	3.0	

```
#matches_rolling = matches_rolling.droplevel('team') --> makes it difficult to work with name, so we drop this extra index levels
# we want unique values in our index
matches_rolling.index = range(matches_rolling.shape[0])
matches_rolling
```

	date	time	comp	round	day	venue	result	gf	ga	opponent	...
0	2020-10-04	14:00	Premier League	Matchweek 4	Sun	Home	W	2.0	1.0	Sheffield Utd	...
1	2020-10-17	17:30	Premier League	Matchweek 5	Sat	Away	L	0.0	1.0	Manchester City	...
2	2020-10-25	19:15	Premier League	Matchweek 6	Sun	Home	L	0.0	1.0	Leicester City	...
3	2020-11-01	16:30	Premier League	Matchweek 7	Sun	Away	W	1.0	0.0	Manchester Utd	...
4	2020-11-08	19:15	Premier League	Matchweek 8	Sun	Home	L	0.0	3.0	Aston Villa	...
...
1312	2022-03-13	14:00	Premier League	Matchweek 29	Sun	Away	W	1.0	0.0	Everton	...
1313	2022-03-18	20:00	Premier League	Matchweek 30	Fri	Home	L	2.0	3.0	Leeds United	...

```
#RETRAINING OUR ML MODEL
def make_predictions(data, predictors):
    train= data[data["date"] < '2022-01-01']
    test = data[data["date"] > '2022-01-01']
    rf.fit(train[predictors], train["target"])
    preds = rf.predict(test[predictors])
    combined = pd.DataFrame(dict(actual=test["target"], predicted=preds), index=test.index)
    precision = precision_score(test["target"], preds)
    return combined, precision
combined, precision = make_predictions(matches_rolling, predictors + new_cols)
precision

0.625

combined #can't really see if we are mispredicting any teams results particularly
```

	actual	predicted
55	0	0
56	1	0
57	1	0
58	1	1
59	1	1
...
1312	1	0

So we are merging team, date , opponent and result based on index

```
combined = combined.merge(matches_rolling[["date", "team", "opponent", "result"]],left_index=True, right_index=True)
combined
```

	actual	predicted	date	team	opponent	result
55	0	0	2022-01-23	Arsenal	Burnley	D
56	1	0	2022-02-10	Arsenal	Wolves	W
57	1	0	2022-02-19	Arsenal	Brentford	W
58	1	1	2022-02-24	Arsenal	Wolves	W
59	1	1	2022-03-06	Arsenal	Watford	W
...
1312	1	0	2022-03-13	Wolverhampton Wanderers	Everton	W
1313	0	0	2022-03-18	Wolverhampton Wanderers	Leeds United	L
1314	1	0	2022-04-02	Wolverhampton Wanderers	Aston Villa	W
1315	0	0	2022-04-08	Wolverhampton Wanderers	Newcastle Utd	L
1316	0	0	2022-04-24	Wolverhampton Wanderers	Burnley	L

276 rows × 6 columns

#COMBINING HOME AND AWAY PREDICTIONS

#before combining them we have to normalize the name columns

```
class MissingDict(dict): # Pandas Math dictionary by default will remove missing name and all its data but we would create a mapping dictio
```

```
    __missing__ = lambda self, key: key
```

```
map_values = {
```

```
    "Brighton and Hove Albion" : "Brighton",
```

```
    "Manchester United": "Manchester Utd",
```

```
    "Newcastle United": "Newcastle Utd",
```

```
    "West Ham United": "West Ham",
```

```
    "Wolverhampton Wanderers": "Wolves"
```

```
}
```

```
mapping = MissingDict(**map_values)
```

```
mapping["Wolverhampton Wanderers"]
```

```
    'Wolves'
```

```
combined["new_team"] = combined["team"].map(mapping)
```

```
combined
```



```
merged = combined.merge(combined, left_on=["date", "new_team"], right_on=["date", "opponent"]) #look for new team field and merge that with t
merged
```

	actual	predicted	date	team	opponent	result	new_team
55	0	0	2022-01-23	Arsenal	Burnley	D	Arsenal
2022-							
actual_x	predicted_x	date	team_x	opponent_x	result_x	new_team_x	
0	0	0	2022-01-23	Arsenal	Burnley	D	Arsenal
1	1	0	2022-02-10	Arsenal	Wolves	W	Arsenal
2	1	0	2022-02-19	Arsenal	Brentford	W	Arsenal
3	1	1	2022-02-24	Arsenal	Wolves	W	Arsenal
4	1	1	2022-03-06	Arsenal	Watford	W	Arsenal
...
242	1	0	2022-03-13	Wolverhampton Wanderers	Everton	W	Wolves
243	0	0	2022-03-18	Wolverhampton Wanderers	Leeds United	L	Wolves

```
merged[(merged["predicted_x"] == 1) & (merged["predicted_y"] == 0)][actual_x].value_counts() # showing only those results where algorithm h
```

```
1    24
0    12
Name: actual_x, dtype: int64
```

```
24/36 #Precision Rate
0.6666666666666666
```