Actividad Módulo 16: Data Massaging 1

Para el archivo fifa_eda.csv:

- 1. Mostrar las primeras 20 filas del archivo, las últimas 5 y un sample de 10
- 2. Generar data estadística con .describe() y además los tipos de datos del dataset
- 3. Si es necesario, pasar a numéricas por lo menos 2 columnas que contengan números para incluirlas en el .describe()
- 4. Añadir una columna 'Years Playing' que calcule el año actual menos la columna 'Joined'
- 5. Buscar y mostrar a todos los jugadores de México
- 6. Ordenar y mostrar los datos por la columna Release Clause (sueldo)
- 7. Generar un nuevo dataset que contenga el año y el número de jugadores
- 8. Opcional: Generar un gráfico que contenga, por año, el númerod de jugadores

```
In [ ]: # Importar Librerías y archivo
import pandas as pd

df = pd.read_csv('d:/Documentos/DataAnalysis/EBAC/Python/Modulo14/fifa_eda.csv')
# 1.1 - Primeras 20 filas del df
df.head(20)
```

Out[]:	ID Name		Age	Nationality	Overall	Potential	Club	Value	Wage	Prefei F	
	0	158023	L. Messi	31	Argentina	94	94	FC Barcelona	110500.0	565.0	
	1	20801	Cristiano Ronaldo	33	Portugal	94	94	Juventus	77000.0	405.0	R
	2	190871	Neymar Jr	26	Brazil	92	93	Paris Saint- Germain	118500.0	290.0	R
	3	193080	De Gea	27	Spain	91	93	Manchester United	72000.0	260.0	R
	4	192985	K. De Bruyne	27	Belgium	91	92	Manchester City	102000.0	355.0	R
	5	183277	E. Hazard	27	Belgium	91	91	Chelsea	93000.0	340.0	R
	6	177003	L. Modrić	32	Croatia	91	91	Real Madrid	67000.0	420.0	R
	7	176580	L. Suárez	31	Uruguay	91	91	FC Barcelona	80000.0	455.0	R
	8	155862	Sergio Ramos	32	Spain	91	91	Real Madrid	51000.0	380.0	R
	9	200389	J. Oblak	25	Slovenia	90	93	Atlético Madrid	68000.0	94.0	R
	10	188545	R. Lewandowski	29	Poland	90	90	FC Bayern München	77000.0	205.0	R
	11	182521	T. Kroos	28	Germany	90	90	Real Madrid	76500.0	355.0	R
	12	182493	D. Godín	32	Uruguay	90	90	Atlético Madrid	44000.0	125.0	R
	13	168542	David Silva	32	Spain	90	90	Manchester City	60000.0	285.0	
	14	215914	N. Kanté	27	France	89	90	Chelsea	63000.0	225.0	R
	15	211110	P. Dybala	24	Argentina	89	94	Juventus	89000.0	205.0	
	16	202126	H. Kane	24	England	89	91	Tottenham Hotspur	83500.0	205.0	R
	17	194765	A. Griezmann	27	France	89	90	Atlético Madrid	78000.0	145.0	
	18	192448	M. ter Stegen	26	Germany	89	92	FC Barcelona	58000.0	240.0	R
	19	192119	T. Courtois	26	Belgium	89	90	Real Madrid	53500.0	240.0	
In []:		.2 - ÚL tail(5)	timas 5 filo	as de	L df						

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Out[]:		ID	Name	Age	Nationality	Overall	Potential	Club	Value	Wage	Pref
	18202	238813	J. Lundstram	19	England	47	65	Crewe Alexandra	60.0	1.0	
	18203	243165	N. Christoffersson	19	Sweden	47	63	Trelleborgs FF	60.0	1.0	
	18204	241638	B. Worman	16	England	47	67	Cambridge United	60.0	1.0	
	18205	246268	D. Walker-Rice	17	England	47	66	Tranmere Rovers	60.0	1.0	
	18206	246269	G. Nugent	16	England	46	66	Tranmere Rovers	60.0	1.0	
In []:	# 1.3	- Sampl	e de 10 filas	del	df						

	ID	Name	Age	Nationality	Overall	Potential	Club	Value	Wage	Prefer F
5453	214217	B. Vidal	27	Chile	70	72	CD Palestino	1700.0	6.0	Ri
15891	242721	E. Dick	23	United States	58	65	Sporting Kansas City	130.0	1.0	Ri
5057	239441	Jordi Mboula	19	Spain	70	85	AS Monaco	3800.0	15.0	Ri
3126	190942	K. Hansen	29	Denmark	73	73	FC Midtjylland	3100.0	18.0	Ri
10737	178120	M. Bruccini	32	Italy	65	65	Cosenza	400.0	1.0	Ri
11081	225557	R. Poole	20	Wales	64	75	Manchester United	675.0	13.0	Ri
12346	215112	R. Contreras	22	Argentina	63	71	CD Antofagasta	550.0	1.0	
7817	205094	Jo Young Cheol	29	Korea Republic	67	67	Gyeongnam FC	750.0	3.0	Ri
9145	244305	R. Ramírez	22	Venezuela	66	71	Fortuna Sittard	725.0	2.0	Ri
1558	212729	Tozé	25	Portugal	76	80	Vitória Guimarães	10000.0	11.0	Ri
	15891 5057 3126 10737 11081 12346 7817 9145	5453 214217 15891 242721 5057 239441 3126 190942 10737 178120 12346 215112 7817 205094 9145 244305 1558 212729	5453 214217 B. Vidal 15891 242721 E. Dick 5057 239441 Jordi Mboula 3126 190942 K. Hansen 10737 178120 M. Bruccini 11081 225557 R. Poole 12346 215112 R. Contreras 7817 205094 Jo Young Cheol 9145 244305 R. Ramírez	5453 214217 B. Vidal 27 15891 242721 E. Dick 23 5057 239441 Jordi Mboula 19 3126 190942 K. Hansen 29 10737 178120 M. Bruccini 32 11081 225557 R. Poole 20 12346 215112 R. Contreras 22 7817 205094 Jo Young Cheol 29 9145 244305 R. Ramírez 22	5453 214217 B. Vidal 27 Chile 15891 242721 E. Dick 23 United States 5057 239441 Jordi Mboula 19 Spain 3126 190942 K. Hansen 29 Denmark 10737 178120 M. Bruccini 32 Italy 11081 225557 R. Poole 20 Wales 12346 215112 R. Contreras 22 Argentina 7817 205094 Jo Young Cheol 29 Korea Republic 9145 244305 R. R. R. R. R. R. Ramírez 22 Venezuela	5453 214217 B. Vidal 27 Chile 70 15891 242721 E. Dick 23 United States 58 5057 239441 Jordi Mboula Mboula 19 Spain 70 3126 190942 K. Hansen M. Hansen 29 Denmark 73 10737 178120 M. Bruccini 32 Italy 65 11081 225557 R. Poole 20 Wales 64 12346 215112 R. Contreras 22 Argentina 63 7817 205094 Jo Young Cheol 29 Korea Republic 67 9145 244305 R. Ramírez 22 Venezuela 66	5453 214217 B. Vidal 27 Chile 70 72 15891 242721 E. Dick 23 United States 58 65 5057 239441 Jordi Mboula Mboula 19 Spain 70 85 3126 190942 K. Hansen 29 Denmark 73 73 10737 178120 M. Bruccini 32 Italy 65 65 11081 225557 R. Poole 20 Wales 64 75 12346 215112 R. Poole 20 Argentina 63 71 7817 205094 Jo Young Cheol 29 Korea Republic 67 67 9145 244305 R. Ramírez 22 Venezuela 66 71	5453 214217 B. Vidal 27 Chile 70 72 CD Palestino 15891 242721 E. Dick 23 United States 58 65 Sporting Kansas City 5057 239441 Jordi Mboula Mobula 19 Spain 70 85 AS Monaco 3126 190942 K. Hansen 29 Denmark 73 73 FC Midtjylland 10737 178120 M. Bruccini 32 Italy 65 65 Cosenza 11081 225557 R. Poole 20 Wales 64 75 Manchester United 12346 215112 R. Contreras 22 Argentina 63 71 Antofagasta 7817 205094 Jo Young Cheol 29 Korea Republic 67 67 Gyeongnam Gyeongnam Sittard 9145 244305 R. Ramírez 22 Venezuela 66 71 Fortuna Sittard	5453 214217 B. Vidal 27 Chile 70 72 CD Palestino Palestino 1700.0 15891 242721 E. Dick 23 United States 58 65 Sporting Kansas City 130.0 5057 239441 Jordi Mboula 19 Spain 70 85 AS Monaco 3800.0 3126 190942 K. Hansen 29 Denmark 73 73 FC Midtjylland 3100.0 10737 178120 M. Bruccini 32 Italy 65 65 Cosenza 400.0 11081 225557 R. Poole 20 Wales 64 75 Manchester United 675.0 12346 215112 R. Contreras 22 Argentina 63 71 Antofagasta 550.0 7817 205094 Jo Young Cheol 29 Korea Republic 67 67 Gyeongnam FC 750.0 9145 244305 R. Ramírez 22 Venezuela 66 71<	5453 214217 B. Vidal 27 Chile 70 72 CD Palestino Palestino 1700.0 6.0 15891 242721 E. Dick 23 United States 58 65 Sporting Kansas City 130.0 1.0 5057 239441 Jordi Mboula Mboula 19 Spain 70 85 AS Monaco 3800.0 15.0 3126 190942 K. Hansen 29 Denmark 73 73 FC Midtjylland 3100.0 18.0 10737 178120 M. Bruccini 32 Italy 65 65 Cosenza 400.0 1.0 11081 225557 R. Poole 20 Wales 64 75 Manchester United 675.0 13.0 12346 215112 R. Contreras 22 Argentina 63 71 Antofagasta 550.0 1.0 7817 205094 Jo Young Cheol 29 Korea Republic 67 67 Gyeongnam Sittard 75.0 2.

In []: # 2.1 Estadísticas de describe()
df.describe()

```
Out[]:
                                                                                             Inte
                         ID
                                    Age
                                               Overall
                                                          Potential
                                                                          Value
                                                                                       Wage
                                                                                               R
         count
                18207.000000 18207.000000 18207.000000 18207.000000
                                                                    17955.000000 18207.000000 181!
              214298.338606
                                25.122206
                                                         71.307299
         mean
                                             66.238699
                                                                     2444.530214
                                                                                    9.731312
           std
                29965.244204
                                 4.669943
                                             6.908930
                                                          6.136496
                                                                     5626.715434
                                                                                   21.999290
                   16.000000
                                16.000000
                                            46.000000
                                                         48.000000
                                                                       10.000000
                                                                                    0.000000
          min
          25%
               200315.500000
                                21.000000
                                             62.000000
                                                         67.000000
                                                                      325.000000
                                                                                    1.000000
          50%
               221759.000000
                                25.000000
                                             66.000000
                                                         71.000000
                                                                      700.000000
                                                                                     3.000000
          75%
              236529.500000
                                            71.000000
                                                                                    9.000000
                                28.000000
                                                         75.000000
                                                                     2100.000000
          max 246620.000000
                                45.000000
                                             94.000000
                                                         95.000000 118500.000000
                                                                                   565.000000
In []: # 2.2 Tipo de datos del df
        df.info()
        # A todos los campos numéricos del df les fue asignado el tipo correcto de valor
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 18207 entries, 0 to 18206
        Data columns (total 18 columns):
              Column
                                         Non-Null Count Dtype
              ----
         0
              ID
                                         18207 non-null int64
          1
             Name
                                         18207 non-null object
          2
                                         18207 non-null int64
             Age
          3
             Nationality
                                         18207 non-null object
          4
             0verall
                                         18207 non-null int64
          5
             Potential
                                         18207 non-null int64
         6
             Club
                                         17966 non-null object
                                         17955 non-null float64
         7
             Value
         8
             Wage
                                         18207 non-null float64
          9
             Preferred Foot
                                         18207 non-null object
          10 International Reputation 18159 non-null float64
          11 Skill Moves
                                         18159 non-null float64
          12 Position
                                         18207 non-null object
         13 Joined
                                         18207 non-null int64
          14 Contract Valid Until
                                         17918 non-null object
          15 Height
                                         18207 non-null float64
         16 Weight
                                         18207 non-null float64
          17 Release Clause
                                         18207 non-null float64
         dtypes: float64(7), int64(5), object(6)
        memory usage: 2.5+ MB
        # 4 - Añadir columna Years Playing
        df['Years Playing'] = 2022 - df['Joined']
        df
```

Out[]:		ID	Name	Age	Nationality	Overall	Potential	Club	Value	Wage
	0	158023	L. Messi	31	Argentina	94	94	FC Barcelona	110500.0	565.0
	1	20801	Cristiano Ronaldo	33	Portugal	94	94	Juventus	77000.0	405.0
	2	190871	Neymar Jr	26	Brazil	92	93	Paris Saint- Germain	118500.0	290.0
	3	193080	De Gea	27	Spain	91	93	Manchester United	72000.0	260.0
	4	192985	K. De Bruyne	27	Belgium	91	92	Manchester City	102000.0	355.0
	•••									
	18202	238813	J. Lundstram	19	England	47	65	Crewe Alexandra	60.0	1.0
	18203	243165	N. Christoffersson	19	Sweden	47	63	Trelleborgs FF	60.0	1.0
	18204	241638	B. Worman	16	England	47	67	Cambridge United	60.0	1.0
	18205	246268	D. Walker-Rice	17	England	47	66	Tranmere Rovers	60.0	1.0
	18206	246269	G. Nugent	16	England	46	66	Tranmere Rovers	60.0	1.0

18207 rows × 19 columns

```
In [ ]: # 5 - Mostrar a todos los jugadores de México
df.loc[df['Nationality'] == 'Mexico']
```

Out[]:		ID	Name	Age	Nationality	Overall	Potential	Club	Value	Wage	Preferre Foo
	306	171897	A. Guardado	31	Mexico	82	82	Real Betis	19000.0	35.0	Le
	329	221992	H. Lozano	22	Mexico	81	86	PSV	24000.0	22.0	Rigl
	371	193165	J. Corona	25	Mexico	81	83	FC Porto	21500.0	18.0	Rigl
	397	169416	C. Vela	29	Mexico	81	81	Los Angeles FC	17500.0	15.0	Le
	406	156519	H. Herrera	28	Mexico	81	81	FC Porto	17500.0	20.0	Rigl
	•••										
	17373	244828	R. Gutiérrez	22	Mexico	54	64	Monarcas Morelia	80.0	1.0	Rigl
	17614	239739	A. Ocejo	20	Mexico	53	62	Santos Laguna	90.0	2.0	Le
	18037	246089	C. Landa	19	Mexico	50	60	Tiburones Rojos de Veracruz	50.0	1.0	Le
	18068	240286	J. García	20	Mexico	50	62	Santos Laguna	40.0	1.0	Rigl
	18113	237045	R. Pasquel	22	Mexico	50	60	Deportivo Toluca	40.0	2.0	Rigl
	366 rov	us x 19 c	columns								

366 rows × 19 columns

```
In [ ]: # 6 - Ordernar datos por La columna 'Release Clause'
df.sort_values(by='Release Clause')
```

Out[]:		ID	Name	Age	Nationality	Overall	Potential	Club	Value	Wage
	16204	176860	C. Månsson	37	Sweden	58	58	Kristiansund BK	10.0	1.0
	17209	237751	Y. Nishibe	37	Japan	55	55	Shimizu S-Pulse	10.0	1.0
	16249	176631	B. Lekström	37	Sweden	58	58	Hammarby IF	10.0	1.0
	16565	232893	K. Kitamoto	36	Japan	57	57	Vissel Kobe	10.0	1.0
	16628	102881	K. Stamatopoulos	38	Canada	57	57	AIK	10.0	1.0
	•••									
	25	231747	K. Mbappé	19	France	88	95	Paris Saint- Germain	81000.0	100.0
	5	183277	E. Hazard	27	Belgium	91	91	Chelsea	93000.0	340.0
	4	192985	K. De Bruyne	27	Belgium	91	92	Manchester City	102000.0	355.0
	0	158023	L. Messi	31	Argentina	94	94	FC Barcelona	110500.0	565.0
	2	190871	Neymar Jr	26	Brazil	92	93	Paris Saint- Germain	118500.0	290.0

18207 rows × 19 columns

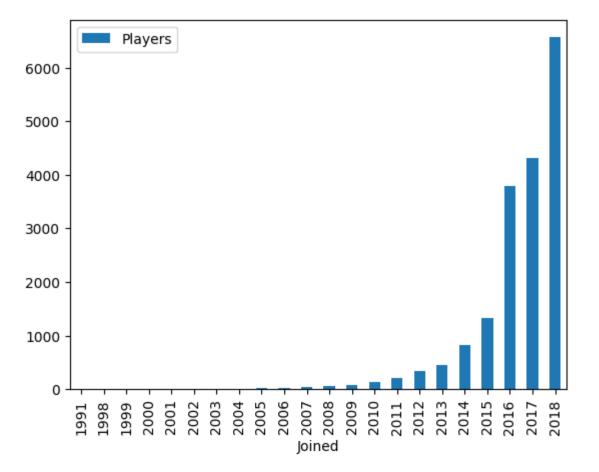
```
In [ ]: # 7 - Generar un dataset que contenga el año (joined) y número de jugadores
df_new = df.groupby(by='Joined').size().rename('Players').reset_index()
df_new
```

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Out[]:		Joined	Players
	0	1991	1
	1	1998	3
	2	1999	1
	3	2000	2
	4	2001	2
	5	2002	10
	6	2003	13
	7	2004	12
	8	2005	17
	9	2006	18
	10	2007	38
	11	2008	53
	12	2009	78
	13	2010	131
	14	2011	201
	15	2012	340
	16	2013	458
	17	2014	818
	18	2015	1336
	19	2016	3799
	20	2017	4307
	21	2018	6569

```
In [ ]: # 8 - Generar un gráfico que contenga el año y número de jugadores
df_new.plot.bar(x='Joined', y='Players')
```

Out[]: <AxesSubplot: xlabel='Joined'>



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