```
In [ ]: from azure.storage.blob import BlobServiceClient, generate_blob_sas, BlobSasPerm
        import pandas as pd
        import json
        import math
        Conexão com a plataforma Azure Blob Storage
In [ ]: #enter credentials
        account name = "filmesteste"
        account_access_key = "S5wta6QAtuV6yYubHeMSen4m09CHuP3A0kIYyZmEpEWj6uUugxXzYdrpMu
        container_name = "filmesbd"
In [ ]: #create a client to interact with blob storage
        connect_str = 'DefaultEndpointsProtocol=https;AccountName=' + account_name + ';A
        blob_service_client = BlobServiceClient.from_connection_string(connect_str)
In [ ]: #use the client to connect to the container
        container_client = blob_service_client.get_container_client(container_name)
In [ ]: def get_data(file_name):
            blob_client = container_client.get_blob_client(file_name)
            stream_downloader = blob_client.download_blob()
            stream = json.loads(stream_downloader.readall())
            return pd.DataFrame(stream)
In [ ]: blob_list = []
        for blob_i in container_client.list_blobs():
            blob_list.append(blob_i.name)
        print(blob list)
       ['actors.json', 'directors.json', 'movies.json', 'producers.json', 'ratings.jso
       n', 'tmdb 5000 movies.json']
        EXTRAÇÃO DE DADOS
In [ ]: bases = {}
        for nome_arquivo in blob_list:
            bases[nome_arquivo] = get_data(nome_arquivo)
In [ ]: #redefinindo nome da base de dados utilizadas no trabalho e definidas no catálog
        df movies = bases['movies.json']
        df_ratings = bases['ratings.json']
        df_box_office = bases['tmdb_5000_movies.json']
        df_actors = bases['actors.json']
        df_directors = bases['directors.json']
        df_producers = bases['producers.json']
        TRANSFORMAÇÃO DE DADOS
In [ ]: | n_movies_without_actors = len(df_movies[df_movies.all_actors.isna()])
        n_movies_without_directors = len(df_movies[df_movies.director.isna()])
        n_movies_without_producers = len(df_movies[df_movies.producer.isna()])
        print('FILMES SEM LISTA DE ATORES: ' + str(n_movies_without_actors))
```

```
print('FILMES SEM LISTA DE DIRETORES: ' + str(n_movies_without_directors))
        print('FILMES SEM LISTA DE PRODUTORES: ' + str(n_movies_without_producers))
       FILMES SEM LISTA DE ATORES: 321
       FILMES SEM LISTA DE DIRETORES: 267
       FILMES SEM LISTA DE PRODUTORES: 930
In [ ]: #REMOVE TODOS OS FILMES OS QUAIS NÃO POSSUEM OS ATORES, DIRETORES E PRODUTORES Q
        df_movies = df_movies[df_movies.all_actors.notna()]
        df_movies = df_movies[df_movies.director.notna()]
        df_movies = df_movies[df_movies.producer.notna()]
In [ ]: #Redefina o índice do DataFrame e use o padrão.
        #inplace = True : Se deve modificar o DataFrame em vez de criar um novo.
        #drop = True :This resets the index to the default integer index.
        df_movies.reset_index(inplace=True, drop=True)
In [ ]: df_movies.info()
       <class 'pandas.core.frame.DataFrame'>
       RangeIndex: 14238 entries, 0 to 14237
       Data columns (total 9 columns):
        # Column
                           Non-Null Count Dtype
       --- -----
                            -----
           title
        0
                            14238 non-null object
                           14238 non-null int64
        1
          year
        2 genre 14238 non-null object
3 _id 14238 non-null object
4 director 14238 non-null object
5 all_actors 14238 non-null object
        6 gender_percent 14182 non-null float64
            adjusted_budget 5435 non-null float64
        7
            producer
                             14238 non-null object
       dtypes: float64(2), int64(1), object(6)
       memory usage: 1001.2+ KB
        CONTABILIZANDO A QUANTIDADE DE ATORES E ATRIZES
In [ ]: df_movies_final = pd.DataFrame()
        for index, movie in df movies.iterrows(): #Iterar sobre linhas do DataFrame como
            male count = 0
            female count = 0
            for actor in movie.all actors:
                 if actor['gender'] == 'male':
                    male count += 1
                     female_count +=1
            movie['actors_count'] = male_count
            movie['actresses_count'] = female_count
            df_movies_final = pd.concat([df_movies_final, pd.DataFrame([movie])], ignore
        CONTABILIZANDO A QUANTIDADE DE PRODUTORES HOMENS E MULHERES
In [ ]: df_movies_final_1 = pd.DataFrame()
        for index, movie in df_movies_final.iterrows(): #Iterar sobre linhas do DataFran
```

CONTABILIZANDO A QUANTIDADE DE DIRETORES HOMENS E MULHERES

```
In []: df_movies_final_2 = pd.DataFrame()
    for index, movie in df_movies_final_1.iterrows(): #Iterar sobre linhas do DataFr
        male_count = 0
        female_count = 0
        for director in movie.director:

        if director['gender'] == 'male':
            male_count += 1
        else:
            female_count += 1

        movie['director_count'] = male_count
        movie['directress_count'] = female_count

        df_movies_final_2 = pd.concat([df_movies_final_2, pd.DataFrame([movie])], ig
```

REMOVE VARIÁVEIS MULTIVALORADAS

```
In [ ]: df_movies_final_2.drop(['director', 'producer', 'all_actors'], axis= 1 ,inplace=
#df_movies_final_2.drop(['all_actors'], axis= 1 ,inplace= True )
```

CHECA SE AS VARIÁVEIS CRIADAS DE FATO ESTÃO PRESENTES NO DATASET

```
In [ ]: df_ratings
```

Out[]:		tconst	averageRating	numVotes
	0	tt0000001	5.7	1993
	1	tt0000002	5.8	268
	2	tt0000003	6.5	1879
	3	tt0000004	5.5	177
	4	tt0000005	6.2	2662
	•••			
	1352356	tt9916730	8.3	10
	1352357	tt9916766	7.0	22
	1352358	tt9916778	7.2	36
	1352359	tt9916840	8.8	6
	1352360	tt9916880	8.2	6

1352361 rows × 3 columns

Renomeando a coluna para ter o mesmo nome da coluna de df_movies_final_2

In []: df_ratings.rename(columns={'tconst': '_id'}, inplace=True)

Junção dos dados de averageRating e numVotes a df_movies_final_2

In []: df_movies_final_2 = df_movies_final_2.merge(df_ratings, on='_id')

In []: df_movies_final_2.head()

Out[]:		title	year	genre	_id	gender_percent	adjusted_budg
	0	In_Old_Madrid_(1911)	1911	[Comedy, Romance, Short]	tt0001695	50.0	Ni
	1	Stick_Around_(1925)	1925	[Comedy, Short]	tt0005864	25.0	Na
	2	Calling_DrPorky_(1940)	1940	[Animation, Comedy, Family, Short]	tt0032298	50.0	Ni
	3	The_Chewin'_Bruin_(1940)	1940	[Animation, Comedy, Family, Short]	tt0032331	0.0	Ni
	4	A_House_Divided_(1913/I)	1913	[Comedy, Short]	tt0002983	50.0	Ni
	4						•

TRATAMENTO DE DADOS BOX OFFICE

df box office.head()

FORMATA NOME DOS FILMES PARA PADRONIZAR DA MESMA FORMA QUE OS NOMES DE FILMES PRESENTES NO DATAFRAME MOVIES

```
Out[]:
                budget
                                genres
                                                                         homepage
                                                                                           id keyword
                                                                                                    [{"id
                              [{"id": 28,
                                                                                                    146
                               "name":
                                                                                                  "name
             237000000
                              "Action"},
                                                        http://www.avatarmovie.com/
                                                                                        19995
                                                                                                  "cultur
                              {"id": 12,
                                                                                                  clash"
                                "nam...
                                                                                                   {"id":
                                                                                                    [{"id
                              [{"id": 12,
                                                                                                     270
                               "name":
                                                                                                  "name
                                                                                          285
             300000000
                                          http://disney.go.com/disneypictures/pirates/
                          "Adventure"},
                                                                                                 "ocean"
                           {"id": 14, "...
                                                                                                {"id": 72
                                                                                                     "na
                                                                                                    [{"id
                              [{"id": 28,
                                                                                                     470
                               "name":
                                                                                                  "name
             245000000
                              "Action"},
                                         http://www.sonypictures.com/movies/spectre/
                                                                                      206647
                                                                                                   "spy"
                              {"id": 12,
                                                                                                {"id": 81
                                "nam...
                                                                                                  "name
                                                                                                    [{"id
                              [{"id": 28,
                                                                                                     849
                               "name":
                                                                                                  "name
             250000000
                              "Action"},
                                                  http://www.thedarkknightrises.com/
                                                                                        49026
                                                                                                      "C
                              {"id": 80,
                                                                                                 comics"
                                                                                                    {"id
                                "nam...
                                                                                                    853,
                                                                                                    [{"id
                              [{"id": 28,
                                                                                                     81
                               "name":
                                                                                                  "name
             260000000
                              "Action"},
                                                 http://movies.disney.com/john-carter
                                                                                       49529
                                                                                               "based o
                              {"id": 12,
                                                                                                  novel"
                                "nam...
                                                                                                   {"id":
         4
         df_box_office['release_date'] = pd.to_datetime(df_box_office.release_date)
         df_box_office = df_box_office[df_box_office.release_date.notna()]
In [ ]:
In [ ]: df_box_office['year'] = df_box_office.release_date.dt.year
          df_box_office['year'] = df_box_office.year.astype(int)
```

```
C:\Users\950604\AppData\Local\Temp\ipykernel_20900\4014192577.py:1: SettingWithCo
       pyWarning:
       A value is trying to be set on a copy of a slice from a DataFrame.
       Try using .loc[row indexer,col indexer] = value instead
       See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
       e/user_guide/indexing.html#returning-a-view-versus-a-copy
         df_box_office['year'] = df_box_office.release_date.dt.year
       C:\Users\950604\AppData\Local\Temp\ipykernel_20900\4014192577.py:2: SettingWithCo
       pyWarning:
       A value is trying to be set on a copy of a slice from a DataFrame.
       Try using .loc[row_indexer,col_indexer] = value instead
       See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
       e/user_guide/indexing.html#returning-a-view-versus-a-copy
        df box office['year'] = df box office.year.astype(int)
        GARANTE QUE NÃO TERÁ PROBLEMA COM CASE SENSITIVE
In [ ]: df movies final 2['title'] = df movies final 2.title.str.lower()
        df box office['title'] = df box office.title.str.lower()
```

```
In []: df_movies_final_2['title'] = df_movies_final_2.title.str.lower()
    df_box_office['title'] = df_box_office.title.str.lower()

C:\Users\950604\AppData\Local\Temp\ipykernel_20900\1218968412.py:2: SettingWithCo
    pyWarning:
    A value is trying to be set on a copy of a slice from a DataFrame.
    Try using .loc[row_indexer,col_indexer] = value instead

See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
    e/user_guide/indexing.html#returning-a-view-versus-a-copy
    df_box_office['title'] = df_box_office.title.str.lower()
```

JUNTA OS DADOS DE BILHETERIA DE BOX OFFICE EM MOVIES

```
df movies consolidated['actors percent'] = df movies consolidated.actors count/(
        df_movies_consolidated['actors_percent'] = df_movies_consolidated['actors_percer
In [ ]: df_movies_consolidated['women_directing_percent'] = df_movies_consolidated.directing_percent'
        df_movies_consolidated['women_directing_percent'] = df_movies_consolidated['women_directing_percent']
        df_movies_consolidated['men_directing_percent'] = df_movies_consolidated.directed
        df_movies_consolidated['men_directing_percent'] = df_movies_consolidated['men_di
        df_movies_consolidated['women_producing_percent'] = df_movies_consolidated.produ
        df_movies_consolidated['women_producing_percent'] = df_movies_consolidated['women_producing_percent']
        df_movies_consolidated['men_producing_percent'] = df_movies_consolidated.produce
        df_movies_consolidated['men_producing_percent'] = df_movies_consolidated['men_producing_percent']
In [ ]: df movies consolidated.info()
       <class 'pandas.core.frame.DataFrame'>
       Index: 1942 entries, 0 to 2481
       Data columns (total 23 columns):
            Column
                                     Non-Null Count Dtype
           ----
       ---
                                     -----
                                     1942 non-null object
        0
           title
        1
           year
                                     1942 non-null int64
                                     1942 non-null object
        2
          genre
           id
                                     1942 non-null
                                                    object
                                                    float64
        4 gender_percent
                                     1942 non-null
                                     1854 non-null
                                                     float64
           adjusted_budget
                                     1942 non-null
                                                     int64
        6 actors count
                                    1942 non-null
                                                    int64
           actresses count
            producer_count_male
                                     1942 non-null
                                                    int64
        8
        9
            producer_count_female
                                     1942 non-null
                                                    int64
        10 director_count
                                     1942 non-null
                                                    int64
        11 directress_count
                                     1942 non-null
                                                     int64
                                     1942 non-null
                                                     float64
        12 averageRating
        13 numVotes
                                     1942 non-null
                                                     int64
        14 budget
                                     1942 non-null
                                                     int64
                                     1942 non-null
                                                     int64
        15 revenue
        16 actress_percent
                                     1942 non-null
                                                     float64
                                     1942 non-null
                                                    float64
        17 delta_percent
                                     1942 non-null
                                                    float64
        18 actors percent
        19 women_directing_percent 1942 non-null
                                                     float64
                                     1942 non-null
                                                     float64
        20 men_directing_percent
        21 women_producing_percent 1942 non-null
                                                     float64
                                                     float64
        22 men_producing_percent
                                     1942 non-null
       dtypes: float64(10), int64(10), object(3)
       memory usage: 364.1+ KB
        COMPARA SE GENDER PERCENT É UMA VARIAVEL A SER USADA NO LUGAR DE
        ACTRESS PERCENT
       df_movies_consolidated[df_movies_consolidated.delta_percent != 0]
```

Out[]:		title	year	genre	_id	gender_percent
	11	the_texas_chain_saw_massacre	1974	[Horror, Thriller]	tt0072271	11.0
	12	harley_davidson_and_the_marlboro_man	1991	[Action, Crime, Drama, Thriller, Western]	tt0102005	29.0
	19	cat_on_a_hot_tin_roof	1958	[Drama]	tt0051459	38.0
	24	topaz	1969	[Thriller]	tt0065112	20.0
	26	the_exorcist	1973	[Drama, Horror]	tt0070047	36.0
	•••					
	2474	old_dogs	2009	[Comedy, Family]	tt0976238	37.0
	2476	yes_man	2008	[Comedy, Romance]	tt1068680	37.0
	2479	red	2010	[Action, Comedy, Thriller]	tt1245526	26.0
	2480	halloween_ii	2009	[Horror]	tt1311067	28.0
	2481	the_next_three_days	2010	[Crime, Drama, Romance, Thriller]	tt1458175	35.0
	964 rov	ws × 23 columns				
	4					>
In []:	df_mov	vies_consolidated.drop([' <mark>_id', 'ad</mark>	justed	_budget',	'delta_per	cent'], axis= 1

In []: df_movies_consolidated

```
Out[]:
                                           title year
                                                              genre gender_percent actors_coul
                                                          [Animation,
                                                              Family,
             0 snow_white_and_the_seven_dwarfs 1937
                                                             Fantasy,
                                                                                 21.0
                                                                                                1
                                                             Musical,
                                                           Romance]
                                                          [Animation,
                                                              Family,
             1
                                       fantasia 1940
                                                                                 14.0
                                                             Fantasy,
                                                              Music]
                                                          [Adventure,
                                                          Animation,
             2
                                      pinocchio 1940
                                                                                 25.0
                                                       Drama, Family,
                                                           Fantasy,...
                                                       [Crime, Drama,
             9
                                 alone_with_her 2006
                                                                                 58.0
                                                             Thriller]
            10
                           an_inconvenient_truth 2006 [Documentary]
                                                                                  0.0
                                                             [Drama,
         2477
                                    the_wrestler 2008
                                                           Romance,
                                                                                 19.0
                                                               Sport]
                   i_hope_they_serve_beer_in_hell
         2478
                                                           [Comedy]
                                                                                 44.0
                                                             [Action,
         2479
                                           red 2010
                                                            Comedy,
                                                                                 26.0
                                                                                                6
                                                             Thriller]
         2480
                                    halloween_ii 2009
                                                             [Horror]
                                                                                 28.0
                                                       [Crime, Drama,
                                                           Romance,
         2481
                             the_next_three_days 2010
                                                                                 35.0
                                                                                                6
                                                             Thriller]
        1942 rows × 20 columns
         CRIA VARIAVEIS DUMMIES PARA GÊNEROS DE FILMES
In [ ]:
        df = df_movies_consolidated.explode('genre', ignore_index=True)
         generos = df.genre.unique()
         generos
In [ ]:
Out[ ]: array(['Animation', 'Family', 'Fantasy', 'Musical', 'Romance', 'Music',
                 'Adventure', 'Drama', 'Crime', 'Thriller', 'Documentary', 'Horror',
                 'Action', 'Western', 'Comedy', 'War', 'Sci-Fi', 'Sport', 'History',
                 'Biography', 'Mystery', 'Film-Noir', 'Adult'], dtype=object)
In [ ]: df_final = pd.DataFrame()
         for index, row in df_movies_consolidated.iterrows():
             for genero in generos:
```

```
row[genero] = 1 if genero in row.genre else 0
             df_final = pd.concat([df_final, pd.DataFrame([row])])
        df_final.drop('genre', axis=1, inplace=True)
In [ ]:
In [ ]:
        df_final
Out[]:
                                         title year gender_percent actors_count actresses_co
            0 snow_white_and_the_seven_dwarfs 1937
                                                                21.0
                                                                               11
                                      fantasia 1940
                                                                14.0
                                                                                6
            2
                                     pinocchio 1940
                                                                25.0
                                                                                9
            9
                                alone_with_her 2006
                                                                58.0
           10
                          an_inconvenient_truth 2006
                                                                 0.0
         2477
                                   the_wrestler 2008
                                                                19.0
                                                                               93
         2478
                   i_hope_they_serve_beer_in_hell 2009
                                                                44.0
                                                                               50
         2479
                                          red 2010
                                                                26.0
                                                                               66
         2480
                                   halloween_ii 2009
                                                                28.0
                                                                               57
         2481
                            the_next_three_days 2010
                                                                35.0
                                                                               65
        1942 rows × 42 columns
        df_final.to_json('df_movies_consolidated.json')
In [ ]:
```