

Challenge Básico: Introducción a PySpark y Koalas

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Introducción:

Este challenge introduce a los estudiantes a la visualización de grandes bases de datos utilizando PySpark y Koalas, un API que facilita el trabajo con grandes volúmenes de datos en un entorno de Spark pero con sintaxis similar a pandas.

Objetivo general:

Familiarizarse con PySpark y la API de Koalas para la manipulación y visualización de datos a gran escala.

Objetivo específico:

Descargar una base de datos desde Kaggle, importarla a Databricks, y realizar un análisis exploratorio utilizando la API pyspark.pandas.

In [1]: # 1. Importación de Bibliotecas

```
import findspark
findspark.init()

import pandas as pd
import pyspark

import pandas as pd
import numpy as np
import pyspark.pandas as ps
from pyspark.sql import SparkSession
import matplotlib as plt
import seaborn as sns
```

```
c:\Users\alfre\anaconda3\envs\ProcessBigData25B\lib\site-packages\pyspark\pandas\__init__.py:50: UserWarning: 'PYARRO_W_IGNORE_TIMEZONE' environment variable was not set. It is required to set this environment variable to '1' in both driver and executor sides if you use pyarrow>=2.0.0. pandas-on-Spark will set it for you but it does not work if there is a Spark context already launched.  
    warnings.warn(
```

```
In [2]: from pyspark.sql import SparkSession  
  
spark = SparkSession.builder\  
    .master("local[*]")\  
    .appName('PySpark_Challenge1')\  
    .getOrCreate()
```

Base de datos:

Base de datos: Global Terrorism Database

Plataforma: Kaggle

Instrucciones: Los estudiantes deberán descargar los datos y cargarlos

Apache Spark

```
In [3]: gt = spark.read.csv('data/globalterrorismdb_0718dist.csv',  
                         inferSchema = True,  
                         header = True)
```

```
In [4]: gt.show(15)
```


197000000001 1970 7 2 NULL 0 NULL 58 Dominican Republic 2 Central America &...
NULL Santo Domingo 18.456792 -69.951164 1 0 NULL NULL NULL NULL NULL NULL
1 1 1 0 NULL NULL NULL 0 1 0 1 1 Assassination
NULL NULL NULL NULL NULL NULL 14 Private Citizens ... 68 Named Civilian
NULL Julio Guzman 58 Dominican Republic NULL NULL NULL NULL NULL NULL NULL NULL N
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197000000002 1970 0 0 NULL 0 NULL 130 Mexico 1 North America
Federal Mexico city 19.371887 -99.086624 1 0 NULL NULL NULL NULL NULL
1 1 1 0 NULL NULL 0 1 0 6 Hostage Taking (K...)
NULL NULL NULL NULL NULL 7 Government (Diplo... 45 Diplomatic Person... Bel
gian Ambassador... Nadine Chaval, da... 21 Belgium NULL NULL NULL NULL NULL NULL NU
LL NULL
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NULL 0 7 NULL NULL NULL NULL NULL NULL NULL NULL NULL
NULL NULL 13 Unknown NULL NULL NULL NULL NULL NULL NULL NULL
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NULL 1 1 0 NULL NULL NULL Mexico 1 800000 NULL NULL
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NULL NULL PGIS 0 1 1
1 NULL
197001000001 1970 1 0 NULL 0 NULL 160 Philippines 5 Southeast Asia
Tarlac Unknown 15.478598 120.599741 4 0 NULL NULL NULL NULL
1 1 1 0 NULL NULL 0 1 0 1 Assassination
NULL NULL NULL NULL NULL 10 Journalists & Media 54 Radio Journalist/...
Voice of America Employee 217 United States NULL NULL NULL NULL NULL NULL NUL
L NULL
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NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL
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NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL
NULL	NULL	PGIS			-9			-9			1
1	NULL										
197001000002	1970	1	0	NULL	0	NULL	78	Greece	8	Western Europe	
Attica	Athens	37.99749	23.762728		1	0		NULL		NULL	
1	1	1	0	NULL	NULL	0	1	0	3	Bombing/Explosion	
NULL	NULL	NULL	NULL	NULL	NULL	7	Government (Diplo...	46	Embassy/Consulate		
NULL	U.S. Embassy	217	United States		NULL		NULL	NULL	NULL	NULL	N
ULL	NULL	NULL	NULL	NULL	NULL		NULL	NULL	NULL	NULL	
Unknown	NULL	NULL	NULL	NULL	NULL		NULL	0	NULL	NULL	
0	NULL	NULL	NULL	NULL	NULL		NULL	NULL	NULL	NULL	
NULL	6	Explosives		16	Unknown Explosive...		NULL		NULL		
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losive	NULL	NULL	NULL	NULL	NULL		1	NULL	NULL	NULL	
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NULL	NULL	PGIS			-9			-9			1
1	NULL										
197001000003	1970	1	0	NULL	0	NULL	101	Japan	4	East Asia	
Fukouka	Fukouka	33.580412	130.396361		1	0		NULL		NULL	
1	1	1	-9	NULL	NULL	0	1	0	7	Facility/Infrastr...	
NULL	NULL	NULL	NULL	NULL	NULL	7	Government (Diplo...	46	Embassy/Consulate		
NULL	U.S. Consulate	217	United States		NULL		NULL	NULL	NULL	NULL	N
ULL	NULL	NULL	NULL	NULL	NULL		NULL	NULL	NULL	NULL	
Unknown	NULL	NULL	NULL	NULL	NULL		NULL	0	NULL	NULL	
0	NULL	NULL	NULL	NULL	NULL		NULL	NULL	NULL	NULL	
NULL	8	Incendiary		NULL	NULL		NULL	NULL	NULL		
NULL	NULL	NULL	NULL	NULL	NULL		NULL		NULL		Ince
ndiary	NULL	NULL	NULL	NULL	NULL		1	NULL	NULL	NULL	
NULL	0	NULL	NULL	NULL	NULL		NULL	0	NULL	NULL	
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NULL	NULL	PGIS			-9			-9			1
1	NULL										
197001010002	1970	1	1	NULL	0	NULL	217	United States	1	North America	
Illinois	Cairo	37.005105	-89.176269		1	0		NULL	1/1/1970: Unknown...		
1	1	1	0	NULL	NULL	0	1	0	2	Armed Assault	
NULL	NULL	NULL	NULL	NULL	NULL	3	Police		22	Police Building (... Cai	
ro Police Depa...	Cairo Police Head...		217	United States		NULL		NULL		NULL	NU
LL	NULL	NULL	NULL	NULL	NULL		NULL		NULL	NULL	

NULL	Black Nationalists	NULL	NULL	NULL	NULL	NULL	NULL	To protest the Ca...	0	NULL
NULL	0	-99	-99	0	NULL	NULL	NULL	NULL	NULL	NULL
NULL	NULL	5	Firearms	5	Unknown Gun Type	NULL	NULL	NULL	NULL	NULL
NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL Seve
ral gunshots ...	0	0	0	0	0	0	1	3 Minor (likely < \$...	NULL	
NULL	0	NULL	NULL	NULL	NULL	NULL	0	NULL	NULL	NULL
NULL	NULL	NULL	NULL	NULL	NULL	NULL	The Cairo Chief o...	""Police Chief Q...	"" Washingt	on Post
on Post	January 2			1970.	""Cairo Police C...		"" Afro-American		January 10	
1970." "Christopher Hewi...										
197001020001	1970	1	2	NULL	0	NULL	218	Uruguay	3	South America
Montevideo	Montevideo	-34.891151	-56.187214		1	0		NULL		NULL
1	1	1	0	NULL	NULL	0	0	0	1	Assassination
NULL	NULL	NULL	NULL	NULL	3	Police		25 Police Security F...		
Uruguayan Police	Juan Maria de Luc...		218		Uruguay	NULL	NULL	NULL	NULL	NUL
L	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL
NULL	Tupamaros (Uruguay)		NULL	NULL	NULL	NULL	NULL	NULL	0	NULL
NULL	0	3	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL
NULL	NULL	5	Firearms		2 Automatic or Semi...		NULL	NULL	NULL	NULL
NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	A
utomatic firearm	0	NULL	NULL	0	NULL	NULL	0	NULL	NULL	NULL
NULL	0	NULL	NULL	NULL	NULL	NULL	0	NULL	NULL	NULL
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NULL	NULL	PGIS			0		0		0	
0	NULL									
197001020002	1970	1	2	NULL	0	NULL	217	United States	1	North America
California	Oakland	37.791927	-122.225906		1	0		Edes Substation	1/2/1970: Unknown...	
1	1	1	2	Other Crime Type	0	1	0	3	Bombing/Explosion	
NULL	NULL	NULL	NULL	NULL	21	Utilities		107	Electricity Pac	
ific Gas & Ele...	Edes Substation		217	United States	NULL	NULL	NULL	NULL	NULL	NU
LL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL
NULL	Unknown		NULL	NULL	NULL	NULL	NULL	NULL	0	NULL
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rmer...	0	NULL	NULL	NULL	NULL	NULL	NULL	0	NULL	NULL
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Civil	and Criminal Dis...	"" U.S. Governmen...			August 6		1970.	Christopher Hewi...	"" Pra	
ege...		2005.								
197001020003	1970	1	2	NULL	0	NULL	217	United States	1	North America
Wisconsin	Madison	43.076592	-89.412488		1	0		NULL	1/2/1970: Karl Ar...	
1	1	1	0	NULL		NULL	0	1	0	7 Facility/Infrastr...

NULL	NULL	NULL	NULL	4	Military	28 Military Recruiti...
R.O.T.C. R.O.T.C. offices ...	217	United States	NULL	NULL	NULL	NULL NULL
NULL	NULL	NULL	NULL	NULL	NULL	NULL
New Year's Gang	NULL	NULL	NULL	NULL To protest the Wa...	0	NULL
0	1	1	1	Letter	NULL	NULL
NULL	8	Incendiary	19	Molotov Cocktail/...	NULL	NULL
NULL	NULL	NULL	NULL	NULL	NULL	NULL Firebomb consi...
sti...	0	0	0	0	0	1
courts...	0	NULL	NULL	NULL	NULL	0
NULL	NULL	NULL	NULL	NULL	The New Years Gan...	"Tom Bates, ""Rad... "" HarperCollinsP...
1979." "The Wisconsin Ca...	1992." "David Newman, Sa...	Heard and Won't ...	"" Mother Jones	February-March		
197001030001 1970	1	3	NULL	0	NULL	217 United States 1 North America
Wisconsin	Madison	43.07295	-89.386694	1	0	NULL 1/3/1970: Karl Ar...
1	1	1	0	NULL	NULL	0
NULL	NULL	NULL	NULL	2 Government (General)	21 Government Buildi...	
Selective Service Selective Service...	217	United States	NULL	NULL	NULL	NU
LL	NULL	NULL	NULL	NULL	NULL	NULL
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NULL	0	1	1	0	NULL	NULL
NULL	NULL	8	Incendiary	20 Gasoline or Alcohol	NULL	NULL
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ed gasoline o...	0	0	0	0	0	1
Slight damage	0	NULL	NULL	NULL	NULL	0
	NULL	NULL	NULL	NULL	NULL	NULL
Civil and Criminal Dis... " U.S. Governmen...	August 6	1970." "Tom Bates, ""Rad... "" Har...	perCollinsP... 1992."			
197001050001 1970	1	1	NULL	0	NULL	217 United States 1 North America
Wisconsin	Baraboo	43.4685	-89.744299	1	0	NULL
1	1	0	1	1 Insurgency/Gueril...	0	0
NULL	NULL	NULL	NULL	4	Military	27 Military Barracks...
NULL	Badger Army ammo ...	217	United States	NULL	NULL	NULL NULL N
ULL	NULL	NULL	NULL	NULL	NULL	NULL Weath...
er Undergrou...	NULL	NULL	NULL	NULL	NULL	0
0	NULL	NULL	NULL	NULL	NULL	NULL
NULL	6	Explosives	16 Unknown Explosive...	NULL	NULL	NULL
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losive	0	NULL	NULL	0	NULL	3 Minor (likely < \$...
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NULL	NULL	NULL	PGIS	0	0	0
0	NULL					

197001060001 1970 1 6 NULL 0 NULL 217 United States 1 North America
Colorado Denver 39.758968 -104.876305 1 0 NULL 1/6/1970: Unknown...
1 1 1 1 2 Other Crime Type 0 1 0 7 Facility/Infrastr...
NULL NULL NULL NULL NULL 4 Military 28 Military Recruiti... Arm
y Recruiting S... Army Recruiting S... 217 United States NULL NULL NULL NULL NU
LL NULL
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Molotov cocktail 0 0 0 0 0 0 1 3 Minor (likely < \$... 305
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Civil and Criminal Dis... " U.S. Governmen... August 6 1970." "Christopher Hewi... "" Pra
ege Securi... 2005."
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Lazio Rome 41.890961 12.490069 1 0 NULL NULL
1 1 1 -9 NULL NULL 0 1 0 4 Hijacking
NULL NULL NULL NULL NULL 6 Airports & Aircraft 42 Aircraft (not at ... Tr
ans World Airline Flight 802 Boeing... 217 United States NULL NULL NULL NULL NU
LL NULL
NULL Unknown NULL NULL NULL NULL NULL NULL NULL NULL 0 NULL
NULL 0 1 NULL
NULL NULL 5 Firearms 4 Rifle/Shotgun (no... 5 Firearms 3
Handgun NULL
file - carbine; ... 0 0 0 0 0 0 0 0 NULL NULL NULL NUL
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197001090001 1970 1 9 NULL 0 NULL 217 United States 1 North America
Michigan Detroit 42.331685 -83.047924 1 0 NULL 1/9/1970: Unknown...
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S. Government h... Packard Propertie... 217 United States NULL NULL NULL NULL NU
ULL NULL
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NULL NULL 8 Incendiary 19 Molotov Cocktail/... NULL NULL NULL NULL
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Firebomb 0 0 0 0 0 0 1 3 Minor (likely < \$... NULL Building
was damaged 0 NULL NULL NULL NULL NULL 0 NULL NULL NULL NULL

NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	"Committee on Gov...
Civil	and Criminal Dis...	"" U.S. Governmen...		August 6			1970."	"Christopher Hewi... "" Pra
ege...		2005."						
197001090002	1970	1	9	NULL	0	NULL	217	United States
Puerto Rico	Rio Piedras	18.386932	-66.061127		1	0 Caparra Shopping ...		1/9/1970: The Ar...
1	1	1	2	Other Crime Type	0	1	0	7 Facility/Infrastr...
NULL	NULL	NULL	NULL	NULL	1	Business		7 Retail/Grocery/Ba... Ame
rican owned bu...	Baker's Store		217	United States	NULL	NULL	NULL	NULL
LL	NULL	NULL	NULL	NULL	NULL	NULL	NULL	NULL
NULL Armed Commandos o...		NULL	NULL	NULL	NULL	NULL To protest United...		1
NULL	0	-99	-99	1	10	Unknown	NULL	NULL
NULL	NULL	8	Incendiary		18	Arson/Fire	NULL	NULL
NULL	NULL	NULL	NULL		NULL	NULL	NULL	NULL Fire
set in back ...	0	0	0	0	0	0	1	2 Major (likely >= ...
Store destroyed		0	NULL	NULL	NULL	NULL	NULL	2000000
LL	NULL	NULL	NULL		NULL	NULL The fire began at... "Committee on the... ""		NU
U.S. Governmen...		August 6			1970." ""No Evidence Of... "" The Virgin Isl...			Janua
ry 13	1970." ""Toward People'...							

only showing top 15 rows

```
In [5]: # Convertir a pandas-on-Spark DataFrame
df = gt.toPandas()
```

```
In [6]: # Mostrar Las primeras filas del DataFrame
```

```
df.head()
```

Out[6]:	eventid	iyear	imonth	iday	approxdate	extended	resolution	country	country_txt	region	...	addnotes	scite1	scite2
0	197000000001	1970	7	2	None	0	None	58	Dominican Republic	2	...	None	None	N
1	197000000002	1970	0	0	None	0	None	130	Mexico	1	...	None	None	N
2	197001000001	1970	1	0	None	0	None	160	Philippines	5	...	None	None	N
3	197001000002	1970	1	0	None	0	None	78	Greece	8	...	None	None	N
4	197001000003	1970	1	0	None	0	None	101	Japan	4	...	None	None	N

5 rows × 135 columns

5 rows x 135 columns

```
In [7]: # Obtener los nombres de las columnas como una lista
```

```
column_headers_list = df.columns.tolist()  
print(column_headers_list)
```

```
['eventid', 'iyear', 'imonth', 'iday', 'approxdate', 'extended', 'resolution', 'country', 'country_txt', 'region', 'region_txt', 'provstate', 'city', 'latitude', 'longitude', 'specificity', 'vicinity', 'location', 'summary', 'crit1', 'crit2', 'crit3', 'doubtterr', 'alternative', 'alternative_txt', 'multiple', 'success', 'suicide', 'attacktype1', 'attacktype1_txt', 'attacktype2', 'attacktype2_txt', 'attacktype3', 'attacktype3_txt', 'targtype1', 'targtype1_txt', 'targsubtype1', 'targsubtype1_txt', 'corp1', 'target1', 'natlty1', 'natlty1_txt', 'targtype2', 'targtype2_txt', 'targsubtype2', 'targsubtype2_txt', 'corp2', 'target2', 'natlty2', 'natlty2_txt', 'targtype3', 'targtype3_txt', 'targsubtype3', 'targsubtype3_txt', 'corp3', 'target3', 'natlty3', 'natlty3_txt', 'gname', 'gsubname', 'gname2', 'gsubname2', 'gname3', 'gsubname3', 'motive', 'guncertain1', 'guncertain2', 'guncertain3', 'individual', 'nperps', 'nperpcap', 'claimed', 'claimmode', 'claimmode_txt', 'claim2', 'claimmode2', 'claimmode2_txt', 'claim3', 'claimmode3', 'claimmode3_tx t', 'compclaim', 'weaptype1', 'weaptype1_txt', 'weapsubtype1', 'weapsubtype1_txt', 'weaptype2', 'weaptype2_txt', 'weapsubtype2', 'weapsubtype2_txt', 'weaptype3', 'weaptype3_txt', 'weapsubtype3', 'weapsubtype3_txt', 'weaptype4', 'weaptype4_txt', 'weapsubtype4', 'weapsubtype4_txt', 'weapdetail', 'nkill', 'nkillus', 'nkillter', 'nwound', 'nwoundus', 'nwoundte', 'property', 'propextent', 'propextent_txt', 'propvalue', 'propcomment', 'ishostkid', 'nhostkid', 'nhostkidu s', 'nhours', 'ndays', 'divert', 'kidhijcountry', 'ransom', 'ransomamt', 'ransomamtus', 'ransompaid', 'ransompaidus', 'ransomnote', 'hostkidoutcome', 'hostkidoutcome_txt', 'nreleased', 'addnotes', 'scite1', 'scite2', 'scite3', 'dbsource', 'INT_LOG', 'INT_IDEO', 'INT_MISC', 'INT_ANY', 'related']
```

Limpieza de Datos

```
In [8]: # Conversión de Columnas Numéricas
# Convertimos la columna 'nkill' (y otras numéricas si es necesario) a tipo numérico, manejando posibles errores de coerción.

# Conversión de columnas a tipo numérico el prefijo "n" (nkill, nkillus, nkillter, nwound, nwoundus, nwoundte)

df['nkill'] = ps.to_numeric(df['nkill'], errors='coerce')
df['nkillus'] = ps.to_numeric(df['nkillus'], errors='coerce')
df['nkillter'] = ps.to_numeric(df['nkillter'], errors='coerce')
df['nwound'] = ps.to_numeric(df['nwound'], errors='coerce')
df['nwoundus'] = ps.to_numeric(df['nwoundus'], errors='coerce')
df['nwoundte'] = ps.to_numeric(df['nwoundte'], errors='coerce')
```

```
In [9]: # Verificación de Valores Nulos
# Identificamos columnas con valores nulos y calculamos su porcentaje.

null_counts = df.isnull().sum()
null_percentage = (null_counts / df.shape[0]) * 100
nulls = ps.DataFrame({'nulos': null_counts, 'porcentaje': null_percentage})
nulls=nulls[nulls['nulos'] > 0]
```

Out[9]:

	nulos	porcentaje
approxdate	172452	94.914993
resolution	179471	98.778145
provstate	421	0.231712
city	434	0.238867
latitude	4556	2.507554
longitude	4557	2.508104
specificity	6	0.003302
location	126196	69.456385
summary	66129	36.396409
doubtterr	1	0.000550
alternative	152241	83.791162
alternative_txt	152414	83.886379
multiple	297	0.163464
success	178	0.097969
suicide	85	0.046783
attacktype1	54	0.029721
attacktype1_txt	35	0.019263
attacktype2	174947	96.288204
attacktype2_txt	175111	96.378467
attacktype3	181084	99.665916
attacktype3_txt	181154	99.704443
targtype1	419	0.230611

	nulos	porcentaje
targtype1_txt	263	0.144751
targsubtype1	10503	5.780694
targsubtype1_txt	10451	5.752074
corp1	42517	23.400719
target1	682	0.375363
natlty1	1600	0.880616
natlty1_txt	1584	0.871810
targtype2	170117	93.629844
targtype2_txt	170289	93.724510
targsubtype2	170853	94.034927
targsubtype2_txt	170911	94.066850
corp2	171508	94.395430
target2	170633	93.913843
natlty2	170845	94.030524
natlty2_txt	170853	94.034927
targtype3	180453	99.318623
targtype3_txt	180472	99.329081
targsubtype3	180563	99.379166
targsubtype3_txt	180570	99.383018
corp3	180653	99.428700
target3	180507	99.348344
natlty3	180539	99.365956

	nulos	porcentaje
natlty3_txt	180540	99.366507
gname	487	0.268037
gsubname	175640	96.669620
gname2	179575	98.835385
gsubname2	181441	99.862404
gname3	181302	99.785900
gsubname3	181637	99.970279
motive	131567	72.412503
guncertain1	687	0.378114
guncertain2	179215	98.637247
guncertain3	180967	99.601521
individual	517	0.284549
nperps	71368	39.279876
nperpcap	69737	38.382198
claimed	66283	36.481169
claimmode	162383	89.373167
claimmode_txt	162472	89.422151
claim2	179356	98.714851
claimmode2	180731	99.471630
claimmode2_txt	180873	99.549785
claim3	181275	99.771040
claimmode3	181497	99.893225

	nulos	porcentaje
claimmode3_txt	181521	99.906435
compclaim	176959	97.395578
weaptype1	692	0.380866
weaptype1_txt	436	0.239968
weapsubtype1	20958	11.534969
weapsubtype1_txt	20846	11.473326
weaptype2	167993	92.460826
weaptype2_txt	168199	92.574206
weapsubtype2	169926	93.524721
weapsubtype2_txt	170047	93.591317
weaptype3	179726	98.918493
weaptype3_txt	179767	98.941059
weapsubtype3	179963	99.048935
weapsubtype3_txt	179980	99.058291
weaptype4	181594	99.946613
weaptype4_txt	181598	99.948814
weapsubtype4	181607	99.953768
weapsubtype4_txt	181609	99.954868
weapdetail	68106	37.484520
nkill	11074	6.094963
nkillus	64922	35.732095
nkillter	67239	37.007337

	nulos	porcentaje
nwound	16440	9.048329
nwoundus	64793	35.661095
nwoundte	69183	38.077285
property	50	0.027519
propextent	117295	64.557408
propextent_txt	117429	64.631159
propvalue	142422	78.386932
propcomment	123759	68.115097
ishostkid	542	0.298309
nhostkid	167657	92.275897
nhostkidus	167853	92.383772
nhours	177390	97.632794
ndays	173472	95.476386
divert	181268	99.767187
kidhijcountry	178334	98.152358
ransom	104420	57.471201
ransomamt	180254	99.209097
ransomamtus	181080	99.663715
ransompaid	180886	99.556940
ransompaidus	181117	99.684079
ransomnote	181165	99.710497
hostkidoutcome	170754	93.980439

	nulos	porcentaje
hostkidoutcome_txt	170728	93.966129
nreleased	171300	94.280950
addnotes	153695	84.591422
scite1	66796	36.763516
scite2	70989	39.071280
scite3	74436	40.968457
dbsource	877	0.482688
INT_LOG	26853	14.779488
INT_IDEO	33185	18.264526
INT_MISC	8094	4.454816
INT_ANY	1851	1.018763
related	84106	46.290680

4. Análisis Estadístico Descriptivo

4.1 Estadísticas Generales

```
In [10]: # Estadísticas Descriptivas
# Calculamos estadísticas descriptivas para entender la distribución de las variables numéricas.

df.describe()
```

Out[10]:

	eventid	iyear	imonth	iday	extended	country	region	latitude
count	1.816910e+05	181691.000000	181691.000000	181691.000000	181691.000000	181691.000000	181691.000000	177135.000000
mean	2.002705e+11	2002.638997	6.467277	15.505644	0.045346	131.968501	7.160938	23.498343
std	1.325957e+09	13.259430	3.388303	8.814045	0.208063	112.414535	2.933408	18.569242
min	1.970000e+11	1970.000000	0.000000	0.000000	0.000000	4.000000	1.000000	-53.154613
25%	1.991021e+11	1991.000000	4.000000	8.000000	0.000000	78.000000	5.000000	11.510046
50%	2.009022e+11	2009.000000	6.000000	15.000000	0.000000	98.000000	6.000000	31.467463
75%	2.014081e+11	2014.000000	9.000000	23.000000	0.000000	160.000000	10.000000	34.685087
max	2.017123e+11	2017.000000	12.000000	31.000000	1.000000	1004.000000	12.000000	74.633553



In [11]: # Análisis estadístico descriptivo

```
print(df[['nkill', 'nkillus', 'nkillter', 'nwound', 'nwoundus', 'nwoundte']].describe())
```

	nkill	nkillus	nkillter	nwound	\
count	170617.000000	116769.000000	114452.000000	165251.000000	
mean	2.403823	0.053336	0.509497	3.152641	
std	11.554776	5.728405	4.196603	35.939637	
min	-9.000000	0.000000	0.000000	0.000000	
25%	0.000000	0.000000	0.000000	0.000000	
50%	0.000000	0.000000	0.000000	0.000000	
75%	2.000000	0.000000	0.000000	2.000000	
max	1570.000000	1360.000000	500.000000	8191.000000	
	nwoundus	nwoundte			
count	116898.000000	112508.000000			
mean	0.049556	0.117467			
std	3.145368	1.969353			
min	0.000000	0.000000			
25%	0.000000	0.000000			
50%	0.000000	0.000000			
75%	0.000000	0.000000			
max	751.000000	405.000000			

```
In [12]: # Análisis de Correlación
# Calculamos la correlación entre variables numéricas para observar posibles relaciones.

correlation_matrix = df[['nkill', 'nwound']].corr()
print("Matriz de Correlación entre 'nkill' y 'nwound':\n", correlation_matrix)
```

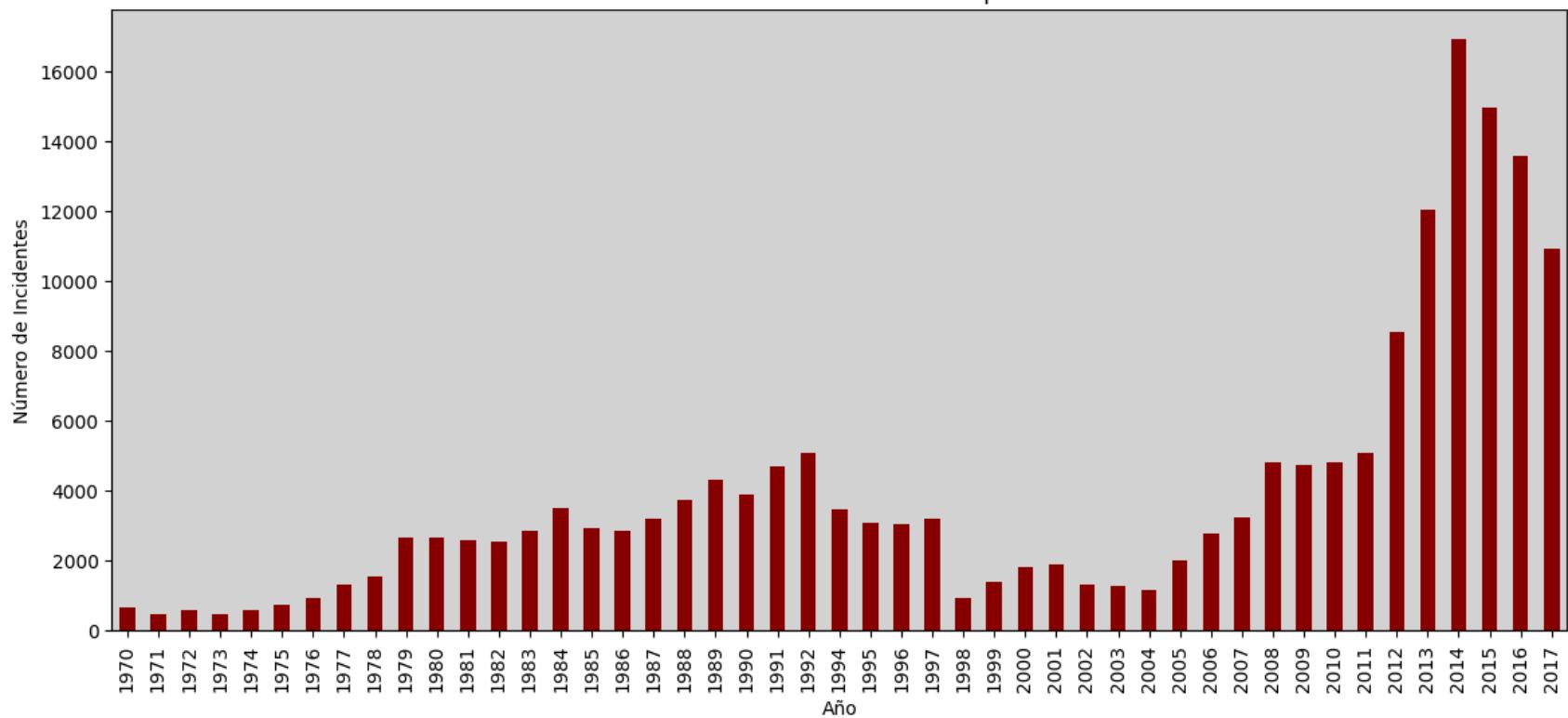
Matriz de Correlación entre 'nkill' y 'nwound':

	nkill	nwound
nkill	1.000000	0.534501
nwound	0.534501	1.000000

```
In [18]: # Análisis Exploratorio de Datos (EDA)
# Incidentes por Año
# Graficamos la cantidad de incidentes terroristas reportados por año.

ax = df['iyear'].value_counts().sort_index().plot(kind='bar', figsize=(14, 6), color='darkred')
ax.set_xlabel("Año")
ax.set_ylabel("Número de Incidentes")
ax.set_title("Número de Incidentes Terroristas por Año")
ax.set_facecolor('lightgrey')
```

Número de Incidentes Terroristas por Año

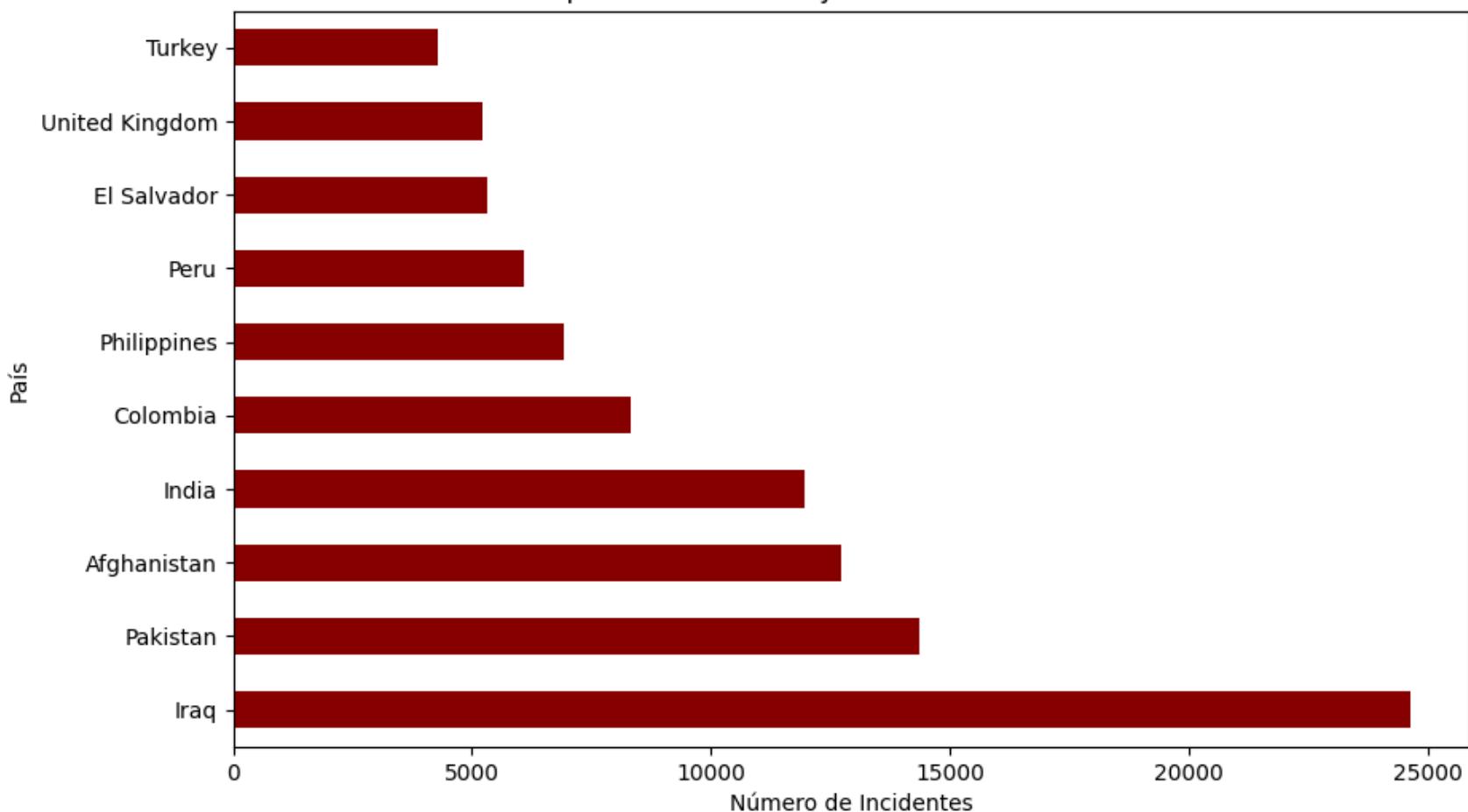


```
In [ ]: # Países con Mayor Número de Incidentes
# Mostramos Los 10 países con mayor número de incidentes.

ax = df['country_txt'].value_counts().head(10).plot(kind='barh', figsize=(10, 6), color='darkred')
ax.set_xlabel("Número de Incidentes")
ax.set_ylabel("País")
ax.set_title("Top 10 Países con Mayor Número de Incidentes")
```

```
Out[ ]: Text(0.5, 1.0, 'Top 10 Países con Mayor Número de Incidentes')
```

Top 10 Países con Mayor Número de Incidentes

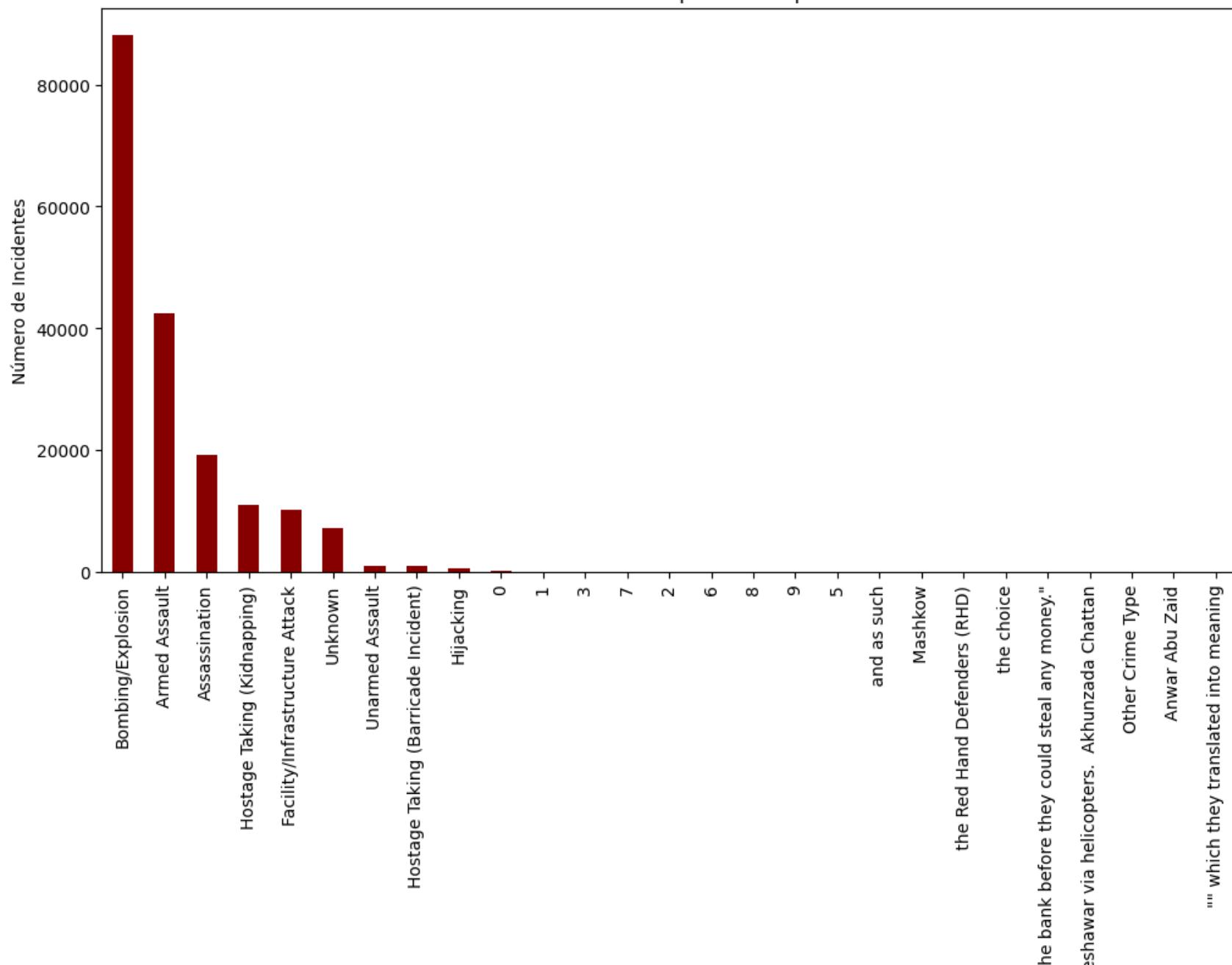


```
In [20]: # Tipos de Ataques
# Analizamos la distribución de tipos de ataques.

ax = df['attacktype1_txt'].value_counts().plot(kind='bar', figsize=(12, 6), color='darkred')
ax.set_xlabel("Tipo de Ataque")
ax.set_ylabel("Número de Incidentes")
ax.set_title("Distribución de Tipos de Ataques")
```

```
Out[20]: Text(0.5, 1.0, 'Distribución de Tipos de Ataques')
```

Distribución de Tipos de Ataques



he did manage to wound one of the robbers. The perpetrators escape have been shifted.

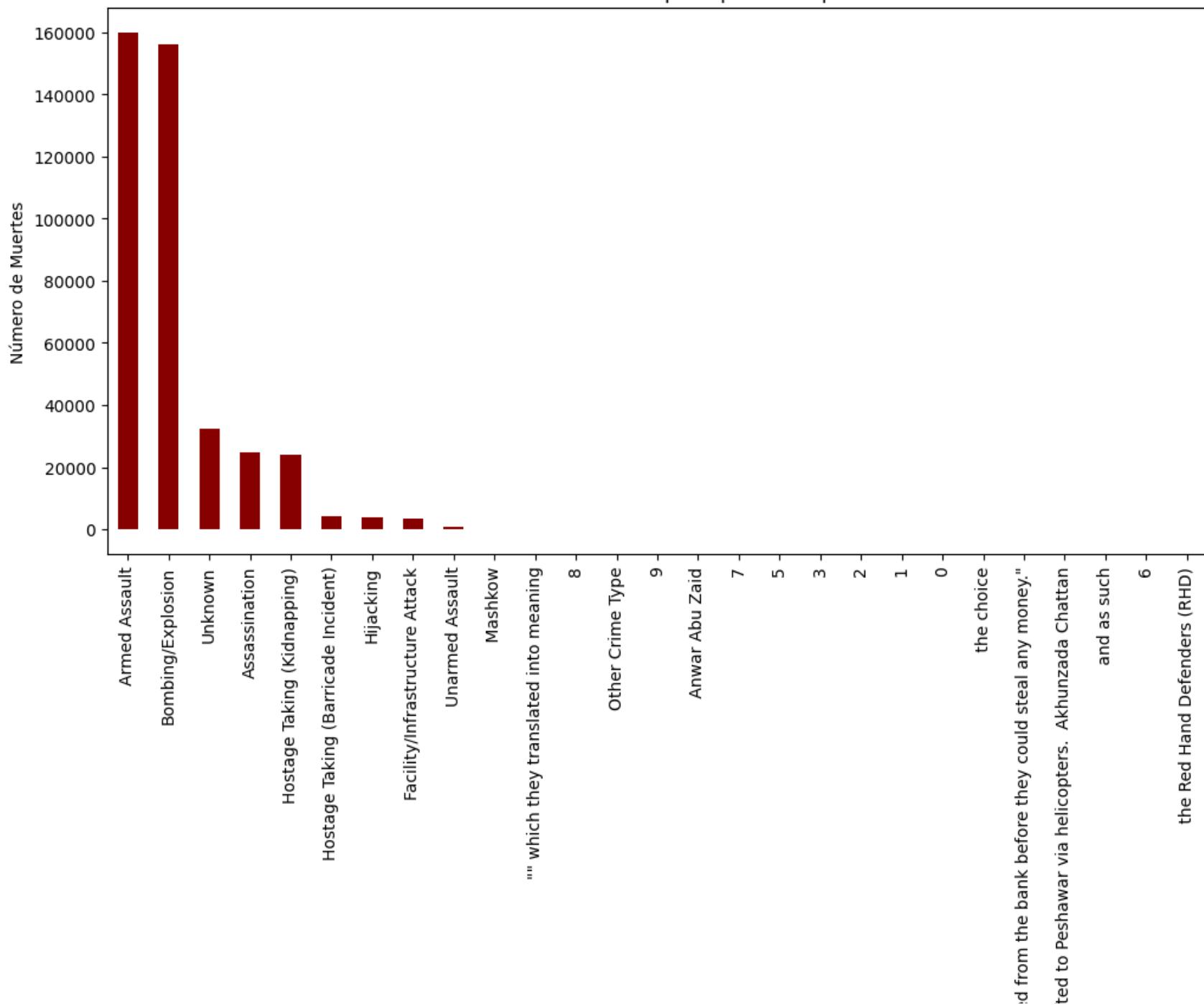
Tipo de Ataque

```
In [21]: # Muertes por Tipo de Ataque
# Exploramos la cantidad total de muertes por cada tipo de ataque.

ax = df.groupby('attacktype1_txt')['nkill'].sum().sort_values(ascending=False).plot(kind='bar', figsize=(12, 6), color='red')
ax.set_xlabel("Tipo de Ataque")
ax.set_ylabel("Número de Muertes")
ax.set_title("Número de Muertes por Tipo de Ataque")
```

```
Out[21]: Text(0.5, 1.0, 'Número de Muertes por Tipo de Ataque')
```

Número de Muertes por Tipo de Ataque



he did manage to wound one of the robbers. The perpetrators escape
have been shif

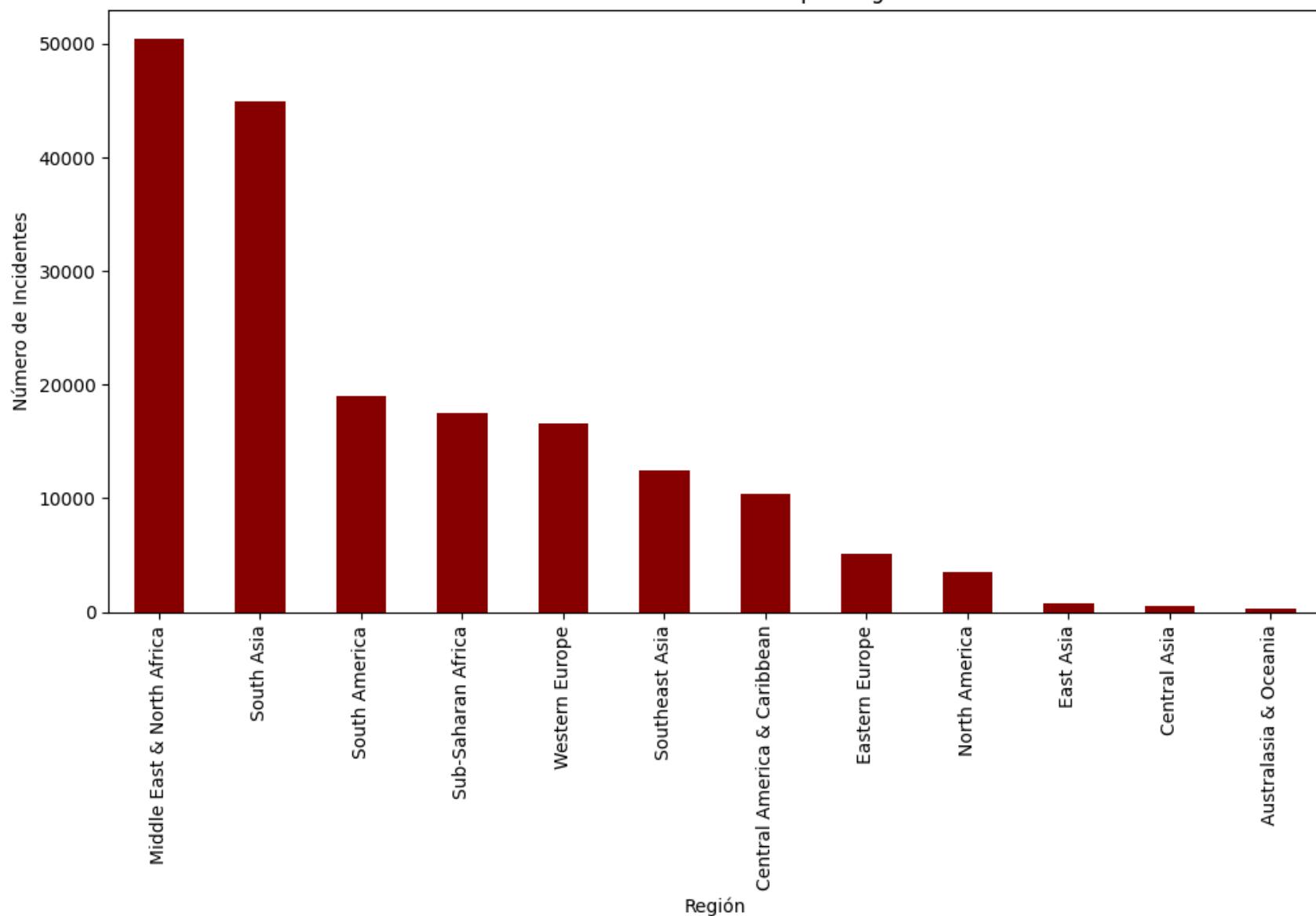
Tipo de Ataque

```
In [22]: # Análisis Geográfico
# Incidentes por Región
# Observamos La cantidad de incidentes en cada región para detectar patrones geográficos.

ax = df['region_txt'].value_counts().plot(kind='bar', figsize=(12, 6), color='darkred')
ax.set_xlabel("Región")
ax.set_ylabel("Número de Incidentes")
ax.set_title("Distribución de Incidentes por Región")
```

```
Out[22]: Text(0.5, 1.0, 'Distribución de Incidentes por Región')
```

Distribución de Incidentes por Región



```
In [23]: # Parar La sesión de Spark  
spark.stop()
```