A tividade SQL

November 6, 2020

1 Atividade Spark SQL

Considerando o dataset detalhado a seguir, extraia o conjunto de informações solicitadas.

1.0.1 Dataset dados da BOVESPA

- Arquivo disponível em /home/dados/bovespa/bovespa.csv
- Dados relativos a bovespa, a bolsa de valores no Brasil
- ~1.3GB
- 8.1M de instâncias

#	Nome do campo	Descrição
0	RegisterType	Fixo '1'
1	TradingDate	Data do pregão
2	BDICode	Utilizado para classificar os papéis na emissão do boletim diário de informações
3	NegociationCode	Codigo de negociação do papel
4	MarketType	Cód. Do mercado em que o papel está cadastrado
5	TradeName	Nome resumido da empresa emissora do papel
6	Specification	Especificação do Papel
7	ForwardMarketTermInDays	Prazo em dias do mercado a termo
8	Currency	Moeda de referência
9	OpeningPrice	Preço de abertura do papel no pregão
10	MaxPrice	Preço máximo do papel no pregão
11	MinPrice	Preço mínimo do papel no pregão
12	MeanPrice	Preço médio do papel no pregão
13	LastTradePrice	Preço do último negócio do papel no pregão
14	BestPurshaseOrderPrice	Preço da melhor oferta de compra do papel no mercado
15	Best Purshase Sale Price	Preço da melhor oferta de venda do papel no mercado
16	NumborOfTrades	Número de negócios efetuados com o papel no pregão
17	${\bf Number Of Traded Stocks}$	Quantidade total de títulos negociados neste papel

#	Nome do campo	Descrição		
18	VolumeOfTradedStocks	Volume total de títulos negociados neste		
		papel		
19	9 PriceForOptionsMarketOrSecondaryTermMa Pre ço de exercício para o mercado de opç			
		ou valor do contrato para o mercado		
20	20 PriceCorrectionsForOptionsMarketOrSecondAmylikadnMacketrreção de preços de exercíc			
		ou valores de contrato		
21	DueDateForOptionsMarketOrSecondaryTermMarketo vencimento para os mercados de			
		opções		
22	FactorOfPaperQuotatuion	Fator de cotação do papel		
23				
		referenciadas em dólar		
24	ISINOrInternCode	Código do papel no sistema ISIN		
25	DistributionNumber	Número de distribuição do papel		

Informações a serem extraídas:

- 1. Quantidade de dias com operações da PETR4 (NegociationCode)
- 2. Maior valor (MaxPrice) histórico por ação (NegociationCode)
- 3. Maior valor (MaxPrice) histórico da PETR4 (NegociationCode)
- 4. Dia ('TradingDate') com a maior quantidade de papeis (NegociationCode) operados
- 5. Dia ('TradingDate') da semana com a maior quantidade de papeis (NegociationCode) operados
- 6. Maior lucro histórico de um papel (NegociationCode) na bovespa (MaxPrice OpeningPrice)
- 7. Maior prejuizo histórico de um papel (NegociationCode) na bovespa (OpeningPrice Last-TradePrice)
- 8. Moeda (Currency) com mais operações
- 9. Papel (NegociationCode) operado em CZ (Currency) com maior quantidade de operações
- 10. Papel (NegociationCode) operado em CZ (Currency) com maior valor médio das operações (MeanPrice)
- 11. Media do preço médio (MeanPrice), mínimo (MinPrice) e máximo (MaxPrice) anual (TradingDate) das ações da PETR4 (NegociationCode)
- 12. Preço médio (MeanPrice) anual (TradingDate) das ações da PETR4 (NegociationCode)
- 13. Preço médio (MeanPrice) anual (TradingDate) das 10 ações (NegociationCode) com mais operações na bovespa
- 14. Desvio Padrão anual do preço médio (MeanPrice) da ação da PETR4 (NegociationCode)
- 15. Desvio Padrão anual do preço médio (MeanPrice) das 10 ações (NegociationCode) com mais operações na bovespa
- 16. Preço médio (MeanPrice) anual (TradingDate) das ações (NegociationCode) com a maior quantidade de operações de acordo com a moeda (Currency)

Dicas: - Crie uma célula (Insert -> Insert Cell Below) para cada informação solicitada - A análise deve ser feita sobre os dados do HDFS - Inicialize o seu cluster executando o script em: Desktop/ambientes/spark/inicializar.sh - Acesse o seu cluster executando o script em: Desktop/ambientes/spark/abrir_navegador.sh

```
[1]: import os
     os.environ['PYSPARK_PYTHON'] = '/usr/bin/python3'
     from pyspark.sql import SparkSession
     sc = SparkSession \
         .builder \
         .master('spark://spark-master:7077') \
         .config('spark.executor.memory', '1g') \
         .getOrCreate()
[2]: df = sc.read \
         .option('delimiter', ',') \
         .option('header', 'true') \
         .option('inferschema', 'true') \
         .csv('hdfs://namenode:9000/bovespa.csv')
[3]: df.printSchema()
    root
     |-- RegisterType: integer (nullable = true)
     |-- TradingDate: integer (nullable = true)
     |-- BDICode: double (nullable = true)
     |-- NegociationCode: string (nullable = true)
     |-- MarketType: integer (nullable = true)
     |-- TradeName: string (nullable = true)
     |-- Specification: string (nullable = true)
     |-- ForwardMarketTermInDays: string (nullable = true)
     |-- Currency: string (nullable = true)
     |-- OpeningPrice: double (nullable = true)
     |-- MaxPrice: double (nullable = true)
     |-- MinPrice: double (nullable = true)
     |-- MeanPrice: double (nullable = true)
     |-- LastTradePrice: double (nullable = true)
     |-- BestPurshaseOrderPrice: double (nullable = true)
     |-- BestPurshaseSalePrice: double (nullable = true)
     |-- NumborOfTrades: double (nullable = true)
     |-- NumberOfTradedStocks: double (nullable = true)
     |-- VolumeOfTradedStocks: double (nullable = true)
     |-- PriceForOptionsMarketOrSecondaryTermMarket: double (nullable = true)
     |-- PriceCorrectionsForOptionsMarketOrSecondaryTermMarket: double (nullable =
    true)
     |-- DueDateForOptionsMarketOrSecondaryTermMarket: double (nullable = true)
     |-- FactorOfPaperQuotatuion: double (nullable = true)
     |-- PointsInPriceForOptionsMarketReferencedInDollarOrSecondaryTerm: double
    (nullable = true)
     |-- ISINOrInternCode: string (nullable = true)
```

```
[4]: df.createOrReplaceTempView('bovespa')
[5]: sc.sql('select NegociationCode from Bovespa').show(4)
   +----+
   |NegociationCode|
   +----+
            ACE 2
            AVI 2|
            CRA 2|
            AGR 11|
   +----+
   only showing top 4 rows
[6]: sc.sql('select NegociationCode, max(MaxPrice) as maximo from bovespa group by
     →NegociationCode').show(10)
   +----+
   |NegociationCode| maximo|
        -----+
            FGO 4| 240000.0|
            VAG 3|5395600.0|
            SUL 4| 310000.0|
            OPM 82|
                      64.01
            OTC 6|1000000.0|
            VSP 4|
                     106.0|
                   40000.01
            ROS 3|
            OTV 85 | 19791.0 |
            OTC 93|
                   15000.0
            OTC 96| 10100.0|
   +----+
   only showing top 10 rows
[7]: sc.sql('select TradingDate, NegociationCode, MaxPrice from bovespa where
     →NegociationCode == \'PETR4\'').show(5)
   +----+
   |TradingDate|NegociationCode|MaxPrice|
   +----+
                      PETR4| 27600.0|
       19980316
       19980317|
                      PETR4| 27800.0|
                      PETR4| 27900.0|
       19980318
       19980319|
                      PETR4| 27700.0|
```

|-- DistributionNumber: double (nullable = true)

```
19980320 | PETR4 | 28000.0 |
   +----+
   only showing top 5 rows
[8]: df.select('TradingDate', 'NegociationCode', 'MaxPrice')\
       .filter(df.NegociationCode == 'PETR4')\
       .orderBy(df.MaxPrice)\
       .show(5)
   +----+
   |TradingDate|NegociationCode|MaxPrice|
   +----+
      20160211
                    PETR4| 427.0|
      201601261
                   PETR4| 431.0|
                   PETR4| 441.0|
      20160210|
      20160120|
                   PETR4| 450.0|
      20160212|
                   PETR4| 451.0|
   +----+
   only showing top 5 rows
[9]: import pyspark.sql.functions as func
   from pyspark.sql.functions import udf
   from pyspark.sql.types import StringType
   udfNegociation2 = udf(lambda negociation: negociation + negociation, u
    →StringType())
   df.select(func.col('NegociationCode'), func.col('MaxPrice'))\
       .filter(func.col('NegociationCode')=='PETR4')\
       .groupBy(func.col('NegociationCode'))\
       .agg(func.max(func.col('MaxPrice')).alias('ValorMaximo'),
          func.count(func.col('MaxPrice').alias('qt')))\
       .withColumn('negociation2', udfNegociation2('NegociationCode'))\
       .show(3)
   +----+
   |NegociationCode|ValorMaximo|count(MaxPrice AS `qt`)|negociation2|
   +----+
                  52100.01
                                        5391 | PETR4PETR4|
           PETR4 |
   +----+
```

[]:

```
[10]: #informacao 1
     sc.sql('select Count(NegociationCode) as QNTDdias from bovespa where
      →NegociationCode==\'PETR4\' ').show(5)
    +----+
     |QNTDdias|
    +----+
         5391
    +----+
[11]: #informacao 2
     sc.sql('select NegociationCode, max(MaxPrice) as PrecoMaximo from bovespa group⊔
      →by NegociationCode order by Negociationcode').show(10)
    +----+
     |NegociationCode|PrecoMaximo|
                       22597.0
             A1LX34
            A1LX34F|
                      22597.0
            AALC11B|
                        3717.0
             AALC34| 10058.0|
            AALC34F| 10058.0|
             AALL34|
                       18737.0
            AALL34F| 19389.0|
                      1976.0
              AALR3|
             AALR3F|
                       1999.0
             AALR3T|
                        1956.0
    only showing top 10 rows
[12]: #informacao 3
     sc.sql('select NegociationCode, max(MaxPrice) as PrecoMaximo from bovespa_{\sqcup}
      →where NegociationCode==\'PETR4\' group by NegociationCode').show()
    +----+
     |NegociationCode|PrecoMaximo|
    +----+
              PETR4|
                      52100.0
    +----+
[13]: #informacao 4
     sc.sql('select TradingDate, count(NegociationCode) as NPapers from bovespa_
      ⇒group by TradingDate order by NPapers desc').show(10)
```

+----+

```
|TradingDate|NPapers|
+----+
            46521
   20191216
   20191212
            4241
   20191213
            42231
   20191118
           4208
   20190819 | 4135 |
   20191114 4131
   20191021
            3975
   20190916
            3954
   20191113|
            3930|
   20190715|
            3916
+----+
only showing top 10 rows
```

```
[14]: #informacao 5
      import datetime
      from pyspark.sql.functions import udf
      from pyspark.sql.types import StringType
      def pegaDiaSemana(x):
          try:
              data = str(x)
              ano, mes, dia = data[0:4], data[4:6], data[6:8]
              return datetime.datetime(int(ano), int(mes), int(dia)).weekday()
          except:
              return '0'
      udfPegaDiaSemana = udf(pegaDiaSemana, StringType())
      df.select('TradingDate')\
          .withColumn('diaSemana', udfPegaDiaSemana('TradingDate'))\
          .groupBy(func.col('diaSemana'))\
          .count()\
          .show()
```

```
+----+
|diaSemana| count|
+-----+
| 3|1597353|
| 0|1626494|
| 5| 3|
| 1|1613747|
| 4|1599650|
| 2|1688444|
```

[15]: #informacao 6

sc.sql('select NegociationCode, max(MaxPrice - OpeningPrice) as lucro from

→bovespa group by NegociationCode order by lucro desc').show(10)

+		-+	+
Negociation	Cod	le	lucro
+		-+	+
l Al	NT	3	4.411E8
l Ai	RN	2	1.5E8
Ti	EL	4	1.0000008E8
l Ri	ΗE	4	1.0E8
l Al	RN	4	9.0E7
LIG	H11	F	8.005E7
l Bi	DL	4	5.0E7
l Al	NT	4	4.0E7
l Co	CC	2	3.322E7
P(OL	3	3.0E7
+		-+	+
			10

only showing top 10 rows

[16]: #informacao 7

sc.sql('select NegociationCode, max(OpeningPrice - LastTradePrice) as prejuizo $_{\cup}$ from bovespa group by NegociationCode order by prejuizo desc').show(10)

+		+-	+
Nego	ociationCod	de	prejuizo
+		+-	+
1	ANT	3	1.45E8
	SLE	2	1.0E8
	RHE	4	5.0E7
	ARN	4	4.0E7
	ANT	4	3.5E7
1	BDL	4	3.0E7
1	LAM	3	2.7E7
1	CCC	2	2.5E7
	BDL	2	2.2E7
1	BMK	3 1	.999999E7
+		+-	+
only	showing to	op 1	0 rows

[17]: #informacao 8

sc.sql('select Currency, Count(Currency) as QNTD from bovespa group by \neg Currency').show()

+----+ |Currency| QNTD|

```
+----+
| CZ$| 458842|
| NCZ$| 172625|
| R$|6995662|
| CR$| 498562|
```

[18]: #informacao 9

sc.sql('select Currency, NegociationCode, Count(NegociationCode) as QNTD from \hookrightarrow bovespa where Currency==\'CZ\$\' group by Currency, NegociationCode order by \hookrightarrow QNTD desc').show(10)

+	+				++
Cur	rency Ne	gocia	atio	onCoo	de QNTD
+	+				++
1	CZ\$			PMA	2 2002
	CZ\$			PET	2 1993
	CZ\$			SHA	2 1755
	CZ\$			BB	2 1751
	CZ\$			BES	2 1749
	CZ\$			FNV	6 1730
	CZ\$			BBD	4 1720
	CZ\$			TRB	2 1679
	CZ\$			AVI	2 1648
	CZ\$			VAG	2 1643
+	+				++
only	showing	top	10	rows	3

[19]: #informacao 10

sc.sql('select Currency, NegociationCode, max(MeanPrice) as MaiorValorMedio \cup from bovespa where Currency==\'CZ\$\' group by Currency, NegociationCode \cup \cup order by MaiorValorMedio desc').show(10)

```
+----+
|Currency|NegociationCode|MaiorValorMedio|
+----+
    CZ$|
                TLS 6|
                           3.7975E7|
                          8000000.01
    CZ$1
                LOB 2|
                ARN 2|
    CZ$|
                          8000000.0|
                CCC 2
    CZ$|
                          7000000.0
    CZ$|
                RHE 4
                          6902439.0|
    CZ$|
                EBC 41
                          6500000.01
    CZ$|
                CON 2
                          6500000.0|
    CZ$1
                AHE 31
                          4000000.01
    CZ$|
                ARN 4
                          4000000.0|
    CZ$|
                MES 1
                          3500000.01
```

```
+----+
only showing top 10 rows
```

```
+----+
|NegociationCode|Year|
                        avg(MinPrice) | avg(MeanPrice) |
    PETR4 | 2019 | 2693.6169354838707 |
                                                2724.5 | 2755.3669354838707 |
         PETR4 | 2018 | 2065.718367346939 | 2102.4489795918366 | 2140.804081632653 |
          PETR4|2017|1441.2032520325204| 1459.060975609756|1477.7642276422764|
         PETR4 | 2016 | 1054.2610441767067 | 1077.4939759036145 | 1101.140562248996 |
          PETR4|2015| 957.8373983739838| 980.4349593495934| 1004.719512195122|
          PETR4 | 2014 | 1636.7701612903227 | 1669.7298387096773 | 1704.1370967741937 |
          PETR4 | 2013 | 1800 . 5725806451612 | 1825 . 3064516129032 | 1852 . 4879032258063 |
          PETR4|2012|2106.9065040650407| 2134.691056910569|2163.5040650406504|
         PETR4 | 2011 | 2344.152610441767 | 2369.9236947791164 | 2396.610441767068 |
         PETR4 | 2010 | 2940.068825910931 | 2975.072874493927 | 3011.076923076923 |
         PETR4 | 2009 | 3141.5853658536585 | 3183.650406504065 | 3225.0243902439024 |
         PETR4|2008| 4705.016064257028| 4806.373493975903| 4906.417670682731|
         PETR4 | 2007 | 5357.355102040817 | 5442.975510204082 | 5522.302040816327 |
         PETR4|2006| 4285.987804878048|4344.0609756097565| 4400.654471544715|
         PETR4|2005| 7993.024096385542| 8094.469879518072| 8192.240963855422|
         PETR4|2004| 8279.132530120482| 8383.638554216868| 8488.269076305221|
          PETR4 | 2003 |
                             5547.0041
                                              5613.9081
                                                                 5681.281
          PETR4|2002|4727.7389558232935|4796.7389558232935| 4871.04016064257|
          PETR4|2001| 5216.276422764227| 5297.70325203252| 5383.422764227642|
          PETR4|2000|23769.959677419356|24186.137096774193| 24586.6935483871|
      ______
```

only showing top 20 rows

```
# Preço médio (MeanPrice) anual (TradingDate) das ações da PETR4□

→ (NegociationCode)

# Vou fazer a média das entradas de PETR4, por exemplo, no ano 2019,

# como [PETR4 ANO Média] para diminuir o número de instâncias Year na tabela.

# Não ficou claro na informação se eu poderia fazer AVG() dos Mean Prices; caso□

→ não seja essa

# a intenção, basta remover Avg() da query e dar group by no mean price também.
```

```
sc.sql('select NegociationCode, SUBSTRING(TradingDate, 1, 4) as Year, 

→Avg(MeanPrice) from bovespa where NegociationCode=\'PETR4\' group by 

→NegociationCode, Year order by Year desc').show(20)
```

```
+----+
|NegociationCode|Year|
                       avg(MeanPrice)|
+----+
          PETR4 | 2019 |
                                 2724.51
          PETR4 | 2018 | 2102.4489795918366 |
          PETR4 | 2017 | 1459.060975609756 |
          PETR4 | 2016 | 1077.4939759036145 |
          PETR4 | 2015 | 980.4349593495934 |
          PETR4 | 2014 | 1669.7298387096773 |
          PETR4 | 2013 | 1825.3064516129032 |
          PETR4 | 2012 | 2134.691056910569 |
          PETR4 | 2011 | 2369.9236947791164 |
          PETR4 | 2010 | 2975.072874493927 |
          PETR4 | 2009 | 3183.650406504065 |
          PETR4|2008| 4806.373493975903|
          PETR4|2007| 5442.975510204082|
          PETR4 | 2006 | 4344.0609756097565 |
          PETR4|2005| 8094.469879518072|
          PETR4|2004| 8383.638554216868|
          PETR4 | 2003 |
                               5613.9081
          PETR4 | 2002 | 4796.7389558232935 |
          PETR4 | 2001 | 5297.70325203252 |
          PETR4|2000|24186.137096774193|
    -----+
only showing top 20 rows
```

```
# Preço médio (MeanPrice) anual (TradingDate) das 10 ações (NegociationCode)

→ com mais operações na bovespa

# Vou fazer a média das entradas de PETR4, por exemplo, no ano 2019,

# como [PETR4 ANO Média] para diminuir o número de instâncias Year na tabela.

# Não ficou claro na informação se eu poderia fazer AVG() dos Mean Prices; caso

→ não seja essa

# a intenção, basta remover Avg() da query e dar group by no mean price também.

top10Acoes = df.select('NegociationCode', 'MeanPrice', 'TradingDate')\
.groupBy('NegociationCode')\
.agg(func.count('NegociationCode').alias('qt'))\
.orderBy(func.desc('qt'))\
.limit(10)\
```

```
.select('Negociationcode').collect()
top10Acoes = [str(i).split('=\'')[1].split('\'')[0] for i in top10Acoes] #__
 \hookrightarrow filtrando.
# Criando uma cláusula where com o array top10:
mywhere3 = ' where '
for i in range(0, len(top10Acoes)):
    mywhere3 += f' NegociationCode==\'{top10Acoes[i]}\' '
    if i < len(top10Acoes)-1:</pre>
         mywhere3 += ' or '
print("Top 10 ações mais negociadas: ", top10Acoes)
sc.sql('select NegociationCode, SUBSTRING(TradingDate, 1, 4) as Year,
 →Avg(MeanPrice) from bovespa'+mywhere3+' group by NegociationCode, Year order⊔
 →by Year desc, NegociationCode').show(50)
Top 10 ações mais negociadas: ['PETR4T', 'VALE5T', 'USIM5T', 'BBAS3T',
'CSNA3T', 'GGBR4T', 'BBDC4T', 'ITSA4T', 'CMIG4T', 'BRKM5T']
+----+
                         avg(MeanPrice)|
|NegociationCode|Year|
          BBAS3T|2019| 4924.428455284553|
          BBDC4T | 2019 | 3643.7828054298643 |
          BRKM5T | 2019 | 3683.0973630831645 |
          CMIG4T | 2019 | 1408.9468302658486 |
          CSNA3T|2019| 1400.867403314917|
          GGBR4T | 2019 | 1466.2689873417721 |
          ITSA4T | 2019 | 1282.3934426229507 |
          PETR4T|2019| 2763.943482224248|
          USIM5T|2019| 880.4004995836802|
          BBAS3T | 2018 | 3565.7605004468273 |
          BBDC4T | 2018 | 3330.918144611187 |
          BRKM5T|2018| 5014.777493606138|
          CMIG4T | 2018 | 871.8730684326711 |
          CSNA3T | 2018 | 920.0129107981221 |
          GGBR4T | 2018 | 1586.2538569424964 |
          ITSA4T | 2018 | 1139.4112820512821 |
          PETR4T | 2018 | 2152.8916467780427 |
          USIM5T|2018| 964.6364551863041|
          BBAS3T | 2017 | 3176.4322508398654 |
          BBDC4T | 2017 | 3229 . 6673684210527 |
          BRKM5T | 2017 | 3974.2268907563025 |
          CMIG4T | 2017 | 849.0871670702179 |
          CSNA3T|2017| 873.8371794871795|
          GGBR4T | 2017 | 1135.5294117647059 |
```

```
ITSA4T | 2017 | 1007.6398713826367 |
          PETR4T | 2017 | 1487.8120950323973 |
          USIM5T|2017| 646.1754601226994|
          VALE5T | 2017 | 2822.725641025641 |
          BBAS3T | 2016 | 2036 . 5633561643835 |
          BBDC4T|2016| 2585.122832369942|
          BRKM5T | 2016 | 2440.7130044843047 |
          CMIG4T | 2016 | 742.6992084432718 |
          CSNA3T | 2016 | 879.3611738148984 |
          GGBR4T | 2016 | 799.6191819464034 |
          ITSA4T|2016| 806.4554896142433|
          PETR4T | 2016 | 1090.7854609929077 |
          USIM5T|2016| 292.8732212160414|
          VALE5T | 2016 | 1441.6931738212527 |
          BBAS3T | 2015 | 2082.96618852459 |
          BBDC4T | 2015 | 2705.7417893544734 |
          BRKM5T | 2015 | 1378.9717514124293 |
          CMIG4T | 2015 | 1002.3052805280528 |
          CSNA3T | 2015 | 519.1433823529412 |
          GGBR4T | 2015 | 767.9319371727748 |
          ITSA4T | 2015 | 842.9115044247787 |
          PETR4T | 2015 | 1002.3142565150741 |
          USIM5T|2015|398.92831105710815|
          VALE5T | 2015 | 1626.9178841309824 |
          BBAS3T | 2014 | 2474.987542468856 |
          BBDC4T | 2014 | 3285.2131147540986 |
     -----
only showing top 50 rows
```

```
[24]: #informacao 14
      #Desvio Padrão anual do preço médio (MeanPrice) da ação da PETR4_
      \rightarrow (NegociationCode)
      sc.sql('select NegociationCode, SUBSTRING(TradingDate, 1, 4) as Year,
       ⇒stddev(MeanPrice) from bovespa where NegociationCode==\'PETR4\' group by⊔
       →NegociationCode, Year order by Year desc').show(20)
```

```
-----+
|NegociationCode|Year|stddev_samp(MeanPrice)|
   -----+
         PETR4 | 2019 |
                      164.42760488118657
         PETR4 | 2018 |
                      322.32991837509013
         PETR4 | 2017 |
                      131.33322761191312
         PETR4 | 2016 |
                         379.52406656319
         PETR4 | 2015 |
                      212.59195810379663
         PETR4 | 2014 |
                       319.8401226321017
         PETR4 | 2013 |
                       136.1536508448088
```

```
PETR4 | 2012 |
                         202.659003035965531
          PETR4 | 2011 |
                         304.10310718438325
          PETR4 | 2010 |
                         392.22804616214563
          PETR4 | 2009 |
                          414.3029746894998|
          PETR4 | 2008 |
                          2311.364312465641
          PETR4 | 2007 |
                          1200.445391820076
          PETR4 | 2006 |
                        242.19193000733335
          PETR4 | 2005 |
                          3377.068887998981
                          757.8230426632923
          PETR4 | 2004 |
          PETR4 | 2003 |
                          863.87485354439331
          PETR4 | 2002 |
                          613.8605232990916
                          333.4576639125791
          PETR4 | 2001 |
          PETR4 | 2000 |
                         20189.956356803512
  -----+
only showing top 20 rows
```

```
[25]: #informacao 15
      # Desvio Padrão anual do preço médio (MeanPrice) das 10 ações (NegociationCode),
      →com mais operações na bovespa
      top10 = df.select('NegociationCode', 'MeanPrice', 'TradingDate')\
          .groupBy('NegociationCode')\
          .agg(func.count('NegociationCode').alias('qt'))\
          .orderBy(func.desc('qt'))\
          .limit(10)
          .select('Negociationcode').collect()
      top10 = [str(i).split('=\'')[1].split('\'')[0] for i in top10] # filtrando.
      # Criando uma cláusula where com o array top10:
      mywhere = ' where '
      for i in range(0, len(top10)):
          mywhere += f' NegociationCode==\'{top10[i]}\' '
          if i < len(top10)-1:
              mywhere += ' or '
      print(top10)
      sc.sql('select NegociationCode, SUBSTRING(TradingDate, 1, 4) as Year, u
      →stddev(MeanPrice) from bovespa'+mywhere+' group by NegociationCode, Year ∪
       →order by Year desc, NegociationCode').show(50)
```

```
['PETR4T', 'VALE5T', 'USIM5T', 'BBAS3T', 'CSNA3T', 'GGBR4T', 'BBDC4T', 'ITSA4T',
'CMIG4T', 'BRKM5T']
+----+
|NegociationCode|Year|stddev_samp(MeanPrice)|
```

		278.4108910417547
	BBDC4T 2019	364.69206520223213
	BRKM5T 2019	889.9095240917367
	CMIG4T 2019	68.05303214710867
	CSNA3T 2019	219.22622899260105
	GGBR4T 2019	143.76969001819543
	ITSA4T 2019	73.34035510641367
	PETR4T 2019	176.76398173897732
	USIM5T 2019	92.63891532765003
	BBAS3T 2018	575.051063933194
	BBDC4T 2018	432.14526038986867
	BRKM5T 2018	443.11194870134796
	CMIG4T 2018	186.25904376536047
	CSNA3T 2018	83.84440052496272
	GGBR4T 2018	99.04875552857983
	ITSA4T 2018	151.4470044266044
	PETR4T 2018	338.21257536293416
	USIM5T 2018	139.37531194931884
	BBAS3T 2017	273.274782356005
	BBDC4T 2017	269.9457573432429
	BRKM5T 2017	597.6399056312631
	CMIG4T 2017	124.77845445317237
	CSNA3T 2017	163.8048806476733
	GGBR4T 2017	112.53931672173698
	ITSA4T 2017	79.64024718269013
	PETR4T 2017	135.1036271323912
	USIM5T 2017	217.04459485048557
	VALE5T 2017	243.00362994900908
	BBAS3T 2016	470.8910488349438
	BBDC4T 2016	433.04996245749703
	BRKM5T 2016	421.8264748924745
	CMIG4T 2016	147.42872993037642
	CSNA3T 2016	236.72543717725782
	GGBR4T 2016	276.12371126963797
	ITSA4T 2016	81.81346489527289
	PETR4T 2016	400.1555589906185
	USIM5T 2016	117.55224599489586
	VALE5T 2016	510.606044374684
	BBAS3T 2015	346.33333266832733
	BBDC4T 2015	535.9770520944436
	BRKM5T 2015	295.61541026880235
1	CMIG4T 2015	288.8073638718649
	CSNA3T 2015	116.06609547866248
	GGBR4T 2015	193.91744668436476
	ITSA4T 2015	111.20937042147659
	PETR4T 2015	213.90144240157417
	USIM5T 2015	118.93182722037069
	VALE5T 2015	282.324250940134

```
[26]: #informacao 16
      #Preço médio (MeanPrice) anual (TradingDate) das ações (NegociationCode) com au
      #quantidade de operações de acordo com a moeda (Currency)
      # Vou fazer a média das entradas de PETR4, por exemplo, no ano 2019,
      # como PETR4 ANO Média para diminuir o tamanho da tabela.
      # Pegando os tickers com mais ocorrências:
      topByCurrency = df.select('NegociationCode', 'Currency')\
          .groupBy('Currency', 'NegociationCode')\
          .agg(func.count('NegociationCode').alias('qt'))\
          .orderBy(func.col('Currency'), func.desc('qt'))\
          .collect()
      # Filtrando:
      topByCurrency = [[str(i).split('=\'')[1].split('\'')[0], str(i).split('=\'')[2].
       →split('\'')[0]] for i in topByCurrency]
      curr_, topMost = [],[]
      for i in range(0, len(topByCurrency)):
          if topByCurrency[i][0] not in curr_:
              curr_.append(topByCurrency[i][0])
              topMost.append(topByCurrency[i])
      print("Top Most By Currency: ", topMost)
      # Criando minha cláusula where:
      mvwhere2 = ' where '
      for i in range(0, len(topMost)):
          mywhere2 += f' NegociationCode=\'{topMost[i][1]}\' '
          if i < len(topMost)-1:</pre>
              mywhere2 += ' or '
      sc.sql('select NegociationCode, SUBSTRING(TradingDate, 1, 4) as Year, __
       →Avg(MeanPrice) as MeanPrice from bovespa'+mywhere2+' group by ⊔
       →NegociationCode, Year order by Year desc, MeanPrice desc').show(50)
```

Top Most By Currency: [['CR\$', 'BBD 4'], ['CZ\$', 'PMA 2'], ['NCZ\$', 'PET 2'],

['R\$', 'PETR4T']]

+	++
NegociationCode Year	MeanPrice ++
PETR4T 2019	2763.943482224248
PETR4T 2018	2152.8916467780427
PETR4T 2017	1487.8120950323973
PETR4T 2016	1090.7854609929077
PETR4T 2015	1002.3142565150741
PETR4T 2014	1671.9566591422122
PETR4T 2013	1849.2040432534086
PETR4T 2012	2154.13648409894
PETR4T 2011	2456.218079493493
PETR4T 2010	2983.363447559709
PETR4T 2009	3264.6164483260554
PETR4T 2008	4897.825913089842
PETR4T 2007	5544.564305500594
PETR4T 2006	4412.334102445778
PETR4T 2005	7806.432778489117
PETR4T 2004	8550.459595959595
PETR4T 2003	5838.4639278557115
PETR4T 2002	4834.294193548387
PETR4T 2001	5357.148730350665
PETR4T 1999	26129.480662983424
PETR4T 1998	20028.90404040404
BBD 4 1998	923.0672268907563
BBD 4 1997	940.3768115942029
BBD 4 1996	908.142378559464
BBD 4 1995	·
BBD 4 1994	
BBD 4 1993	
BBD 4 1992	
PET 2 1991	
BBD 4 1991	
	1.1893549348230913E7
	122173.04329896907
	68709.10914454277
PET 2 1989	
	40317.880952380954
BBD 4 1989	
PET 2 1988	
BBD 4 1988	
PMA 2 1988	
PET 2 1987	
	1507.8791018998272
BBD 4 1987	
PET 2 1986	
BBD 4 1986	1859.125322997416

	PMA 2 1986 1800.8725961538462
	-
[]:	
[]:	