

```
In [1]: # EJERCICIO PRÁCTICA PARA EL ANÁLISIS DEL DATAFRAME DE LA PRODUCCIÓN DE CAFÉ EN C
# ELABORÓ Ing. Luis Armando Amaya Q.
# Se incluyen en este ejercicio práctico algunas instrucciones vistas, puede util
# Recuerde incluir todos los resultados que considere importantes y necesarios
# También la presentación de resultados mediante gráficos
# EL DATAFRAME SE ENCUENTRA EN LA CARPETA del drive en Google creada subcartpeta
# TE LO DEJADO LISTO PARA QUE LO PUEDAS UTILIZAR CON TUS CONOCIMIENTOS Y CREATIVI
# Este ejercicio le permite enriquecer su conocimiento y aportar en la construcci
# IMPORTANTE: Debes DOCUMENTAR CADA LÍNEA CON UN COMENTARIO utilizando el símbolo
# IMPORTANTE: UTILICE SUS CONOCIMIENTOS Y CREATIVIDAD PARA "REALIZAR MINERÍA DE D
# ESTE EJERCICIO ES UNA GUÍA, NO UN INFORME DETALLADO, POR FAVOR, UTILIZARLO COMO
# SOLAMENTE USTED MISMO SE PONE LOS LÍMITES, ASÍ QUE DEBES EXIGIRTE A TI MISMO.
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In [1]: # Recuerda que en INTERNET hay MUCHAS más instrucciones que puedes APLICAR en tu
# Solamente te he mostrado algunas instrucciones importantes, pero hay muchísimas
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In [2]: import pandas as pd
# Importar la librería PANDAS
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In [ ]:
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In [3]: pd.read_csv("PRODUCCIONc.csv")
#Lectura del Dataframe
```

Out[3]:

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
0	2007	ANTIOQUIA	CAFE	112,343.60	120,500.80	1.07	14.54	14.66
1	2007	BOLIVAR	CAFE	502.00	446.00	0.89	0.05	0.07
2	2007	BOYACA	CAFE	11,374.50	9,683.10	0.85	1.17	1.48
3	2007	CALDAS	CAFE	78,393.65	92,815.00	1.18	11.20	10.23
4	2007	CAQUETA	CAFE	2,295.00	2,134.00	0.93	0.26	0.30
...	...	...	...	...	...	...	...	...
261	2018	QUINDIO	CAFE	16,374.73	17,739.03	1.08	2.07	2.21
262	2018	RISARALDA	CAFE	35,874.73	45,918.75	1.28	5.37	4.83
263	2018	SANTANDER	CAFE	42,269.07	55,918.71	1.32	6.53	5.69
264	2018	TOLIMA	CAFE	97,304.04	97,451.31	1.00	11.39	13.11
265	2018	VALLE DEL CAUCA	CAFE	48,305.31	49,667.88	1.03	5.80	6.51

266 rows × 8 columns

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In [4]: produccion_df=pd.read_csv("produccionc.csv")
# Asignación del nombre del Dataframe
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In [5]: produccion_df
# listado general del Dataframe produccion
```

Out[5]:

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
0	2007	ANTIOQUIA	CAFE	112,343.60	120,500.80	1.07	14.54	14.66
1	2007	BOLIVAR	CAFE	502.00	446.00	0.89	0.05	0.07
2	2007	BOYACA	CAFE	11,374.50	9,683.10	0.85	1.17	1.48
3	2007	CALDAS	CAFE	78,393.65	92,815.00	1.18	11.20	10.23
4	2007	CAQUETA	CAFE	2,295.00	2,134.00	0.93	0.26	0.30
...	...	...	...	...	...	...	...	...
261	2018	QUINDIO	CAFE	16,374.73	17,739.03	1.08	2.07	2.21
262	2018	RISARALDA	CAFE	35,874.73	45,918.75	1.28	5.37	4.83
263	2018	SANTANDER	CAFE	42,269.07	55,918.71	1.32	6.53	5.69
264	2018	TOLIMA	CAFE	97,304.04	97,451.31	1.00	11.39	13.11
265	2018	VALLE DEL CAUCA	CAFE	48,305.31	49,667.88	1.03	5.80	6.51

266 rows × 8 columns

```
In [6]: type(produccion_df)
```

Out[6]: pandas.core.frame.DataFrame

```
In [7]: produccion_df.dtypes
```

```
Out[7]: Anio                                int64
Departamento                             object
Producto                                  object
Area (ha)                                 object
Produccion (ton)                          object
Rendimiento (ha/ton)                      float64
Produccion Nacional (ton)                 float64
Area Nacional (ha)                       float64
dtype: object
```

```
In [8]: pd.unique(produccion_df['Anio'])
#indica los valores de los años en el dataframe
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Out[8]: array([2007, 2008, 2009, 2010, 2011, 2012, 2013, 2014, 2015, 2016, 2017,
                2018], dtype=int64)
```

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In [9]: pd.unique(produccion_df['Departamento'])  
#indica los departamentos en el dataframe
```

```
Out[9]: array(['ANTIOQUIA', 'BOLIVAR', 'BOYACA', 'CALDAS', 'CAQUETA', 'CASANARE',  
              'CAUCA', 'CESAR', 'CHOCO', 'CUNDINAMARCA', 'HUILA', 'LA GUAJIRA',  
              'MAGDALENA', 'META', 'NARIÑO', 'NORTE DE SANTANDER', 'PUTUMAYO',  
              'QUINDIO', 'RISARALDA', 'SANTANDER', 'TOLIMA', 'VALLE DEL CAUCA',  
              'ARAUCA', 'GUAVIARE'], dtype=object)
```

```
In [10]: pd.unique(produccion_df['Producto'])  
#indica los productos que están en el dataframe
```

```
Out[10]: array(['CAFE'], dtype=object)
```

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In [11]: pd.unique(produccion_df['Area (ha)'])
```

```
Out[11]: array(['112,343.60', '502.00', '11,374.50', '78,393.65', '2,295.00',
                '2,605.00', '53,471.00', '23,172.00', '290.00', '43,017.30',
                '89,661.56', '4,785.00', '17,506.00', '2,048.00', '24,458.50',
                '30,171.84', '35.00', '19,904.00', '47,689.25', '34,406.67',
                '91,679.10', '76,667.80', '114,694.00', '572.00', '10,778.50',
                '74,897.00', '2,735.00', '2,149.00', '56,208.00', '23,198.00',
                '90.00', '43,633.35', '89,131.20', '4,553.00', '17,521.00',
                '2,146.00', '25,582.00', '31.00', '19,571.00', '47,227.00',
                '34,169.37', '86,829.20', '72,419.00', '112,420.20', '770.00',
                '10,672.50', '73,083.00', '2,332.00', '1,904.00', '57,860.00',
                '23,420.00', '70.00', '43,475.84', '86,726.78', '4,488.00',
                '17,036.00', '2,216.00', '26,467.20', '33,552.58', '23.00',
                '19,052.00', '45,428.00', '37,985.90', '88,667.00', '67,001.30',
                '111,602.71', '0.00', '850.00', '9,427.00', '72,240.58',
                '2,536.00', '2,198.00', '55,162.00', '22,489.50', '157.50',
                '44,264.16', '87,139.53', '4,207.00', '17,000.00', '2,326.00',
                '23,504.05', '30,731.96', '24.00', '18,159.00', '47,308.00',
                '39,000.64', '84,658.70', '69,332.10', '106,419.57', '10.00',
                '8,441.74', '66,331.61', '2,810.00', '2,081.50', '54,246.42',
                '22,350.00', '37,478.87', '78,792.21', '4,100.00', '16,577.00',
                '2,578.00', '24,263.80', '21,520.45', '40.00', '20,139.30',
                '44,733.64', '37,282.04', '93,145.35', '68,038.40', '112,221.14',
                '870.00', '6,698.20', '54,871.88', '2,882.50', '2,322.00',
                '56,825.00', '22,911.00', '37,175.06', '79,809.34', '5,143.00',
                '17,686.00', '2,783.00', '27,806.40', '19,339.31', '42.00',
                '21,109.83', '45,588.03', '33,947.15', '90,904.48', '69,456.71',
                '109,755.50', '659.04', '9,289.05', '60,264.29', '2,905.84',
                '2,232.94', '74,105.64', '25,106.39', '125.01', '36,189.18',
                '118,200.88', '5,750.70', '17,016.72', '2,483.43', '32,136.51',
                '25,332.45', '24.27', '21,203.03', '39,615.60', '38,613.68',
                '97,308.81', '53,481.02', '110,115.86', '936.34', '9,834.39',
                '59,757.18', '3,074.92', '2,599.43', '77,068.46', '26,138.58',
                '136.88', '33,623.54', '128,273.15', '6,078.64', '18,533.11',
                '2,739.71', '33,608.32', '23,724.20', '101.16', '21,462.81',
                '40,154.46', '40,733.20', '100,832.91', '56,035.94', '109,649.61',
                '1,065.07', '10,461.85', '58,376.40', '3,410.56', '2,752.31',
                '77,405.83', '25,948.50', '137.47', '34,101.49', '130,452.40',
                '5,631.53', '17,996.31', '2,922.21', '33,490.93', '22,940.64',
                '128.65', '21,491.21', '41,732.03', '42,679.11', '103,368.73',
                '54,938.79', '105,666.60', '1,065.97', '10,181.80', '56,022.04',
                '3,392.22', '2,671.04', '78,421.95', '25,530.59', '134.96',
                '33,214.17', '126,052.15', '5,531.20', '17,745.80', '2,924.89',
                '32,750.16', '21,520.64', '20,041.70', '40,472.26', '41,387.79',
                '100,328.77', '52,648.25', '99,311.53', '1,137.42', '9,598.33',
                '51,854.59', '3,408.69', '2,436.63', '80,289.56', '25,158.80',
                '125.67', '30,894.16', '122,575.76', '5,340.80', '18,129.50',
                '2,926.85', '33,639.55', '21,409.77', '209.29', '17,699.67',
                '37,334.16', '42,327.26', '96,018.89', '51,470.86', '98,038.15',
                '1,182.13', '9,653.45', '50,762.22', '3,485.24', '2,360.55',
                '82,085.54', '23,915.45', '140.33', '29,085.24', '122,002.46',
                '4,810.97', '17,414.32', '2,761.01', '33,465.54', '20,873.04',
                '209.93', '16,374.73', '35,874.73', '42,269.07', '97,304.04',
                '48,305.31'], dtype=object)
```

```
In [12]: pd.unique(produccion_df['Produccion (ton)'])
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```
Out[12]: array(['120,500.80', '446.00', '9,683.10', '92,815.00', '2,134.00',
                '2,048.40', '51,348.00', '13,278.50', '205.90', '33,729.14',
                '129,052.51', '2,958.70', '14,005.00', '1,617.20', '31,770.05',
                '13,593.24', '34.00', '25,426.00', '72,842.55', '29,469.52',
                '112,322.38', '69,618.24', '113,505.20', '711.00', '9,547.30',
                '86,884.00', '2,469.00', '1,388.13', '48,073.00', '13,841.45',
                '68.00', '78,254.77', '131,316.47', '2,328.90', '14,017.00',
                '1,656.96', '31,262.50', '13,593.25', '35.60', '23,669.00',
                '60,079.00', '29,016.75', '101,201.88', '65,666.43', '103,703.00',
                '292.60', '8,567.97', '81,668.22', '2,332.00', '2,079.70',
                '47,221.00', '12,770.00', '78.75', '37,118.07', '104,609.42',
                '2,340.40', '13,412.80', '1,672.60', '27,487.71', '10,221.69',
                '26.70', '21,985.00', '53,648.00', '26,311.61', '88,633.10',
                '62,711.08', '121,253.38', '0.00', '510.00', '7,083.07',
                '95,957.90', '2,902.50', '2,564.86', '45,113.00', '13,276.08',
                '98.00', '37,214.80', '104,336.56', '2,393.00', '13,600.00',
                '2,221.90', '24,594.10', '22,111.65', '21,065.00', '72,091.00',
                '27,094.16', '94,230.20', '69,496.65', '115,267.98', '12.00',
                '5,643.39', '78,805.87', '2,528.40', '2,023.50', '41,645.39',
                '11,035.85', '32,780.35', '85,150.66', '1,933.00', '13,301.60',
                '2,533.75', '24,073.95', '12,332.00', '45.80', '20,814.11',
                '49,042.31', '22,089.82', '53,288.42', '65,475.63', '91,621.30',
                '652.50', '4,981.59', '54,115.96', '2,446.38', '1,718.25',
                '50,588.14', '19,994.35', '140.00', '30,786.41', '85,212.64',
                '3,434.30', '14,096.05', '2,133.10', '28,077.94', '12,214.54',
                '48.40', '18,030.13', '36,989.43', '23,271.89', '85,027.49',
                '61,190.55', '102,403.24', '395.07', '5,591.05', '58,634.19',
                '2,188.92', '1,338.56', '56,303.92', '15,050.27', '105.93',
                '24,993.74', '115,874.98', '3,447.31', '10,200.84', '1,650.41',
                '28,606.96', '15,185.79', '16.87', '20,599.27', '39,073.92',
                '30,227.02', '77,215.36', '42,948.40', '111,452.91', '606.93',
                '6,364.41', '62,869.38', '2,503.81', '1,688.60', '63,365.76',
                '16,935.63', '125.42', '25,118.55', '135,971.20', '3,923.80',
                '12,012.98', '1,950.84', '32,321.56', '15,108.55', '76.04',
                '22,518.42', '42,719.53', '34,512.79', '86,453.62', '49,799.28',
                '120,365.77', '1,089.74', '9,501.54', '67,231.37', '3,749.27',
                '2,626.73', '83,626.44', '22,240.81', '158.20', '31,165.15',
                '145,168.10', '4,317.50', '16,691.31', '3,206.35', '36,607.56',
                '20,267.64', '124.67', '24,694.56', '47,215.69', '47,304.16',
                '105,563.88', '57,583.56', '119,970.68', '1,128.32', '9,583.80',
                '66,661.14', '3,861.63', '2,638.88', '87,642.49', '22,649.03',
                '160.62', '31,413.34', '145,154.42', '4,387.19', '17,031.09',
                '3,322.42', '37,020.90', '19,590.10', '23,791.30', '47,357.02',
                '47,512.36', '105,976.19', '57,067.08', '140,398.62', '748.97',
                '7,638.99', '68,668.20', '5,108.33', '1,747.51', '97,922.49',
                '16,628.14', '158.85', '33,943.39', '133,787.95', '3,516.80',
                '11,937.90', '4,013.11', '35,004.18', '23,409.44', '282.18',
                '18,792.05', '46,779.71', '54,908.68', '94,556.71', '51,687.80',
                '141,898.91', '734.91', '7,780.34', '68,670.96', '5,280.40',
                '1,629.25', '102,147.00', '14,943.62', '181.42', '32,580.24',
                '136,161.86', '2,990.91', '10,826.24', '3,877.62', '35,679.42',
                '23,471.69', '289.50', '17,739.03', '45,918.75', '55,918.71',
                '97,451.31', '49,667.88'], dtype=object)
```

```
In [13]: pd.unique(produccion_df['Rendimiento (ha/ton)'])
```

```
Out[13]: array([1.07, 0.89, 0.85, 1.18, 0.93, 0.79, 0.96, 0.57, 0.71, 0.78, 1.44,
0.62, 0.8 , 1.3 , 0.45, 0.97, 1.28, 1.53, 0.86, 1.23, 0.91, 0.99,
1.24, 1.16, 0.9 , 0.65, 0.6 , 0.76, 1.79, 1.47, 0.51, 0.77, 1.22,
1.15, 1.21, 1.27, 1.17, 0.92, 0.38, 1.12, 1. , 1.09, 0.82, 0.55,
1.13, 0.52, 0.75, 1.04, 0.3 , 0.69, 0.94, 0. , 1.33, 1.14, 0.59,
0.84, 1.2 , 1.05, 0.72, 1.11, 1.52, 1.08, 0.67, 1.19, 0.49, 0.87,
0.47, 0.98, 1.03, 1.1 , 0.74, 2. , 0.83, 1.01, 0.63, 0.81, 0.88,
0.66, 0.7 , 1.06, 0.64, 1.02, 0.95, 1.41, 1.32, 1.5 , 1.26, 1.37,
1.35, 1.25, 1.45, 1.29, 1.4 , 1.38])
```

```
In [14]: pd.unique(produccion_df['Produccion Nacional (ton)'])
```

```
Out[14]: array([1.454e+01, 5.000e-02, 1.170e+00, 1.120e+01, 2.600e-01, 2.500e-01,
6.190e+00, 1.600e+00, 2.000e-02, 4.070e+00, 1.557e+01, 3.600e-01,
1.690e+00, 2.000e-01, 3.830e+00, 1.640e+00, 0.000e+00, 3.070e+00,
8.790e+00, 3.560e+00, 1.355e+01, 8.400e+00, 1.370e+01, 9.000e-02,
1.150e+00, 1.049e+01, 3.000e-01, 1.700e-01, 5.800e+00, 1.670e+00,
1.000e-02, 9.440e+00, 1.585e+01, 2.800e-01, 3.770e+00, 2.860e+00,
7.250e+00, 3.500e+00, 1.221e+01, 7.930e+00, 1.463e+01, 4.000e-02,
1.210e+00, 1.152e+01, 3.300e-01, 2.900e-01, 6.660e+00, 1.800e+00,
5.240e+00, 1.476e+01, 1.890e+00, 2.400e-01, 3.880e+00, 1.440e+00,
3.100e+00, 7.570e+00, 3.710e+00, 1.250e+01, 8.850e+00, 1.556e+01,
7.000e-02, 9.100e-01, 1.231e+01, 3.700e-01, 5.790e+00, 1.700e+00,
4.780e+00, 1.339e+01, 3.100e-01, 1.750e+00, 3.160e+00, 2.840e+00,
2.700e+00, 9.250e+00, 3.480e+00, 1.209e+01, 8.920e+00, 1.800e+01,
8.000e-02, 8.800e-01, 3.900e-01, 3.200e-01, 6.500e+00, 1.720e+00,
5.120e+00, 1.330e+01, 2.080e+00, 4.000e-01, 3.760e+00, 1.930e+00,
3.250e+00, 7.660e+00, 3.450e+00, 8.320e+00, 1.022e+01, 1.462e+01,
1.000e-01, 7.900e-01, 8.630e+00, 2.700e-01, 8.070e+00, 3.190e+00,
4.910e+00, 1.360e+01, 5.500e-01, 2.250e+00, 3.400e-01, 4.480e+00,
1.950e+00, 2.880e+00, 5.900e+00, 1.357e+01, 9.760e+00, 1.570e+01,
6.000e-02, 8.600e-01, 8.990e+00, 2.100e-01, 2.310e+00, 1.777e+01,
5.300e-01, 1.560e+00, 4.390e+00, 2.330e+00, 5.990e+00, 4.640e+00,
1.184e+01, 6.590e+00, 1.530e+01, 8.700e-01, 2.300e-01, 8.700e+00,
1.867e+01, 5.400e-01, 1.650e+00, 4.440e+00, 2.070e+00, 3.090e+00,
5.860e+00, 4.740e+00, 1.187e+01, 6.840e+00, 1.415e+01, 1.300e-01,
1.120e+00, 7.900e+00, 4.400e-01, 9.830e+00, 2.620e+00, 3.660e+00,
1.707e+01, 5.100e-01, 1.960e+00, 3.800e-01, 4.300e+00, 2.380e+00,
2.900e+00, 5.550e+00, 5.560e+00, 1.241e+01, 6.770e+00, 1.405e+01,
7.810e+00, 4.500e-01, 1.026e+01, 2.650e+00, 3.680e+00, 1.700e+01,
1.990e+00, 4.340e+00, 2.290e+00, 2.790e+00, 6.680e+00, 1.649e+01,
9.000e-01, 8.060e+00, 6.000e-01, 1.150e+01, 3.990e+00, 1.571e+01,
4.100e-01, 1.400e+00, 4.700e-01, 4.110e+00, 2.750e+00, 3.000e-02,
2.210e+00, 5.490e+00, 6.450e+00, 1.110e+01, 6.070e+00, 1.658e+01,
8.020e+00, 6.200e-01, 1.900e-01, 1.194e+01, 3.810e+00, 1.591e+01,
3.500e-01, 1.260e+00, 4.170e+00, 2.740e+00, 5.370e+00, 6.530e+00,
1.139e+01])
```

```
In [15]: pd.unique(produccion_df['Area Nacional (ha)'])
```

```
Out[15]: array([1.466e+01, 7.000e-02, 1.480e+00, 1.023e+01, 3.000e-01, 3.400e-01,
        6.980e+00, 3.020e+00, 4.000e-02, 5.610e+00, 1.170e+01, 6.200e-01,
        2.280e+00, 2.700e-01, 3.190e+00, 3.940e+00, 0.000e+00, 2.600e+00,
        6.220e+00, 4.490e+00, 1.196e+01, 1.000e+01, 1.513e+01, 8.000e-02,
        1.420e+00, 9.880e+00, 3.600e-01, 2.800e-01, 7.410e+00, 3.060e+00,
        1.000e-02, 5.750e+00, 1.175e+01, 6.000e-01, 2.310e+00, 3.370e+00,
        3.980e+00, 2.580e+00, 6.230e+00, 4.510e+00, 1.145e+01, 9.550e+00,
        1.490e+01, 1.000e-01, 1.410e+00, 9.680e+00, 3.100e-01, 2.500e-01,
        7.670e+00, 3.100e+00, 5.760e+00, 1.149e+01, 5.900e-01, 2.260e+00,
        2.900e-01, 3.510e+00, 4.450e+00, 2.520e+00, 6.020e+00, 5.030e+00,
        8.880e+00, 1.499e+01, 1.100e-01, 1.270e+00, 9.710e+00, 2.000e-02,
        5.950e+00, 1.171e+01, 5.700e-01, 3.160e+00, 4.130e+00, 2.440e+00,
        6.360e+00, 5.240e+00, 1.137e+01, 9.310e+00, 1.494e+01, 1.200e-01,
        1.180e+00, 3.900e-01, 7.610e+00, 3.140e+00, 5.260e+00, 1.106e+01,
        5.800e-01, 2.330e+00, 3.410e+00, 2.830e+00, 6.280e+00, 5.230e+00,
        1.308e+01, 1.580e+01, 9.400e-01, 7.720e+00, 4.100e-01, 3.300e-01,
        8.000e+00, 3.220e+00, 1.123e+01, 7.200e-01, 2.490e+00, 3.910e+00,
        2.720e+00, 2.970e+00, 6.420e+00, 4.780e+00, 1.280e+01, 9.780e+00,
        1.422e+01, 9.000e-02, 1.200e+00, 7.810e+00, 3.800e-01, 9.600e+00,
        3.250e+00, 4.690e+00, 1.531e+01, 7.500e-01, 2.200e+00, 3.200e-01,
        4.160e+00, 3.280e+00, 2.750e+00, 5.130e+00, 5.000e+00, 1.261e+01,
        6.930e+00, 1.384e+01, 1.240e+00, 7.510e+00, 9.690e+00, 3.290e+00,
        4.230e+00, 1.612e+01, 7.600e-01, 4.220e+00, 2.980e+00, 2.700e+00,
        5.050e+00, 5.120e+00, 1.267e+01, 7.040e+00, 1.369e+01, 1.300e-01,
        1.310e+00, 7.290e+00, 4.300e-01, 9.660e+00, 3.240e+00, 4.260e+00,
        1.628e+01, 7.000e-01, 2.250e+00, 4.180e+00, 2.860e+00, 2.680e+00,
        5.210e+00, 5.330e+00, 1.290e+01, 6.860e+00, 1.359e+01, 1.400e-01,
        7.200e+00, 4.400e-01, 1.008e+01, 4.270e+00, 1.621e+01, 7.100e-01,
        4.210e+00, 2.770e+00, 5.200e+00, 5.320e+00, 6.770e+00, 1.318e+01,
        1.500e-01, 6.880e+00, 4.500e-01, 1.066e+01, 3.340e+00, 4.100e+00,
        1.627e+01, 2.410e+00, 4.470e+00, 2.840e+00, 3.000e-02, 2.350e+00,
        4.960e+00, 5.620e+00, 1.275e+01, 6.830e+00, 1.321e+01, 1.600e-01,
        1.300e+00, 6.840e+00, 4.700e-01, 3.920e+00, 1.643e+01, 6.500e-01,
        3.700e-01, 2.810e+00, 2.210e+00, 4.830e+00, 5.690e+00, 1.311e+01,
        6.510e+00])
```

```
In [16]: produccion_df['Anio'].min()
```

```
Out[16]: 2007
```

```
In [17]: produccion_df['Anio'].max()
```

```
Out[17]: 2018
```

```
In [18]: produccion_df['Area (ha)'].min()
```

```
Out[18]: '0.00'
```

```
In [20]: produccion_df['Area (ha)'].max()+" Hectarea"
```

```
Out[20]: '99,311.53 Hectarea'
```

```
In [21]: produccion_df['Rendimiento (ha/ton)'].min()
```

```
Out[21]: 0.0
```

```
In [22]: produccion_df['Rendimiento (ha/ton)'].max()
```

```
Out[22]: 2.0
```

```
In [23]: produccion_df['Anio'].isnull()
```

```
Out[23]: 0      False
         1      False
         2      False
         3      False
         4      False
         ...
        261     False
        262     False
        263     False
        264     False
        265     False
        Name: Anio, Length: 266, dtype: bool
```

```
In [24]: produccion_df['Area (ha)'].isnull()
```

```
Out[24]: 0      False
         1      False
         2      False
         3      False
         4      False
         ...
        261     False
        262     False
        263     False
        264     False
        265     False
        Name: Area (ha), Length: 266, dtype: bool
```

```
In [25]: produccion_df['Rendimiento (ha/ton)'].isnull()
```

```
Out[25]: 0      False
         1      False
         2      False
         3      False
         4      False
         ...
        261     False
        262     False
        263     False
        264     False
        265     False
        Name: Rendimiento (ha/ton), Length: 266, dtype: bool
```



```
In [26]: produccion_df['Rendimiento (ha/ton)'].isnull().sum()
```

```
Out[26]: 0
```

```
In [27]: produccion_df['Area (ha)'].isnull().sum()
```

```
Out[27]: 0
```

```
In [28]: produccion_grouped_Anio=produccion_df.groupby("Anio").sum()  
produccion_grouped_Anio
```

```
Out[28]:
```

	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
Anio			
2007	20.91	100.01	100.00
2008	21.62	100.00	99.99
2009	19.39	100.00	99.98
2010	20.84	100.01	100.00
2011	19.65	100.02	100.00
2012	19.75	99.99	100.00
2013	16.71	100.00	99.99
2014	18.09	100.00	100.00
2015	22.54	99.98	100.00
2016	22.34	99.99	100.00
2017	23.50	100.01	100.00
2018	23.75	100.00	100.02

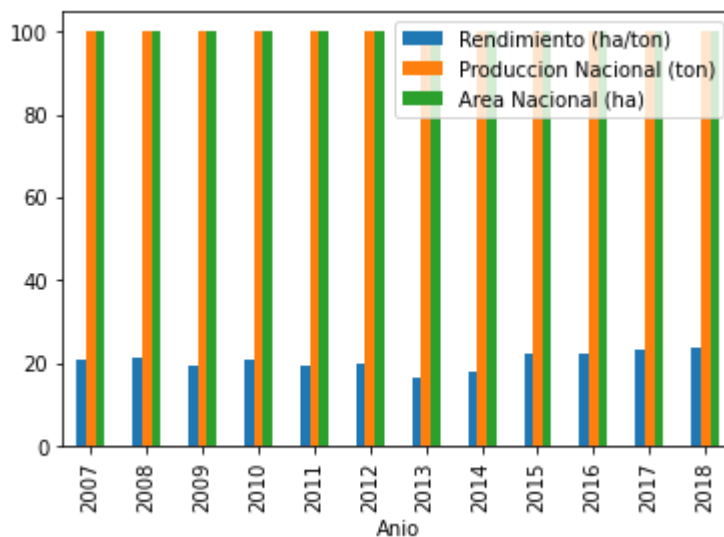
```
In [29]: produccion_grouped_Anio2=produccion_df.groupby("Anio").sum()
produccion_grouped_Anio2
```

Out[29]:

	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
Anio			
2007	20.91	100.01	100.00
2008	21.62	100.00	99.99
2009	19.39	100.00	99.98
2010	20.84	100.01	100.00
2011	19.65	100.02	100.00
2012	19.75	99.99	100.00
2013	16.71	100.00	99.99
2014	18.09	100.00	100.00
2015	22.54	99.98	100.00
2016	22.34	99.99	100.00
2017	23.50	100.01	100.00
2018	23.75	100.00	100.02

```
In [30]: import numpy as np
import re
import sys
%matplotlib inline
produccion_grouped_Anio2.plot(kind='bar')
```

Out[30]: <matplotlib.axes.\_subplots.AxesSubplot at 0x191578af670>



```
In [31]: produccion_grouped_Departamento=produccion_df.groupby(["Anio", "Departamento"]).su
produccion_grouped_Departamento
```

Out[31]:

		Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
Anio	Departamento			
2007	ANTIOQUIA	1.07	14.54	14.66
	BOLIVAR	0.89	0.05	0.07
	BOYACA	0.85	1.17	1.48
	CALDAS	1.18	11.20	10.23
	CAQUETA	0.93	0.26	0.30
...	...	...	...	...
2018	QUINDIO	1.08	2.07	2.21
	RISARALDA	1.28	5.37	4.83
	SANTANDER	1.32	6.53	5.69
	TOLIMA	1.00	11.39	13.11
	VALLE DEL CAUCA	1.03	5.80	6.51

266 rows × 5 columns

```
In [32]: produccion_grouped_Departamento_Rendimiento=produccion_df.groupby(["Anio", "Departamento", "Rendimiento"]).su
produccion_grouped_Departamento_Rendimiento
```

Out[32]:

			Producto	Area (ha)	Produccion (ton)	Produccion Nacional (ton)	Area Nacional (ha)
Anio	Departamento	Rendimiento (ha/ton)					
2007	ANTIOQUIA	1.07	CAFE	112,343.60	120,500.80	14.54	14.66
	BOLIVAR	0.89	CAFE	502.00	446.00	0.05	0.07
	BOYACA	0.85	CAFE	11,374.50	9,683.10	1.17	1.48
	CALDAS	1.18	CAFE	78,393.65	92,815.00	11.20	10.23
	CAQUETA	0.93	CAFE	2,295.00	2,134.00	0.26	0.30
...	...	...	...	...	...	...	...
2018	QUINDIO	1.08	CAFE	16,374.73	17,739.03	2.07	2.21
	RISARALDA	1.28	CAFE	35,874.73	45,918.75	5.37	4.83
	SANTANDER	1.32	CAFE	42,269.07	55,918.71	6.53	5.69
	TOLIMA	1.00	CAFE	11,393.65	13,111.00	11.39	13.11

```
In [53]: produccion_df['Produccion (ton)'].count()  
# cuenta el numero de registros en el dataframe para el campo de La Producción
```

Out[53]: 266

```
In [54]: produccion_df['Anio'].count()  
# cuenta el numero de registros en el dataframe para el campo del año
```

Out[54]: 266

```
In [55]: produccion_grouped_Departamento1=produccion_df.groupby(["Anio","Departamento"]).sum()  
produccion_grouped_Departamento1
```

Out[55]: Rendimiento (ha/ton)            2.00  
Produccion Nacional (ton)        18.67  
Area Nacional (ha)               16.43  
dtype: float64

```
In [56]: produccion_grouped_Departamento=produccion_df.groupby(["Anio","Departamento"]).sum()  
produccion_grouped_Departamento
```

Out[56]: Rendimiento (ha/ton)            0.0  
Produccion Nacional (ton)        0.0  
Area Nacional (ha)               0.0  
dtype: float64

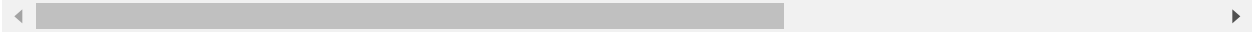
In [57]:

produccion\_grouped\_Departamento=produccion\_df.groupby(["Anio", "Departamento"]).de  
produccion\_grouped\_Departamento

Out[57]:

		Rendimiento (ha/ton)								Produccion Nacional (ton)				
		count	mean	std	min	25%	50%	75%	max	count	mean	...	75%	ma
Anio	Departamento													
2007	ANTIOQUIA	1.0	1.07	NaN	1.07	1.07	1.07	1.07	1.07	1.0	14.54	...	14.54	14.
	BOLIVAR	1.0	0.89	NaN	0.89	0.89	0.89	0.89	0.89	1.0	0.05	...	0.05	0.
	BOYACA	1.0	0.85	NaN	0.85	0.85	0.85	0.85	0.85	1.0	1.17	...	1.17	1.
	CALDAS	1.0	1.18	NaN	1.18	1.18	1.18	1.18	1.18	1.0	11.20	...	11.20	11.
	CAQUETA	1.0	0.93	NaN	0.93	0.93	0.93	0.93	0.93	1.0	0.26	...	0.26	0.
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
2018	QUINDIO	1.0	1.08	NaN	1.08	1.08	1.08	1.08	1.08	1.0	2.07	...	2.07	2.
	RISARALDA	1.0	1.28	NaN	1.28	1.28	1.28	1.28	1.28	1.0	5.37	...	5.37	5.
	SANTANDER	1.0	1.32	NaN	1.32	1.32	1.32	1.32	1.32	1.0	6.53	...	6.53	6.
	TOLIMA	1.0	1.00	NaN	1.00	1.00	1.00	1.00	1.00	1.0	11.39	...	11.39	11.
	VALLE DEL CAUCA	1.0	1.03	NaN	1.03	1.03	1.03	1.03	1.03	1.0	5.80	...	5.80	5.

266 rows × 24 columns



```
In [58]: produccion_Departamento3=produccion_df.groupby(["Anio","Departamento", "Area Nacional"])
produccion_Departamento3
```

Out[58]:

			Rendimiento (ha/ton)								Produccion Nacional		
			count	mean	std	min	25%	50%	75%	max	count	mean	std
Anio	Departamento	Area Nacional (ha)											
2007	ANTIOQUIA	14.66	1.0	1.07	NaN	1.07	1.07	1.07	1.07	1.07	1.0	14.54	NaN
	BOLIVAR	0.07	1.0	0.89	NaN	0.89	0.89	0.89	0.89	0.89	1.0	0.05	NaN
	BOYACA	1.48	1.0	0.85	NaN	0.85	0.85	0.85	0.85	0.85	1.0	1.17	NaN
	CALDAS	10.23	1.0	1.18	NaN	1.18	1.18	1.18	1.18	1.18	1.0	11.20	NaN
	CAQUETA	0.30	1.0	0.93	NaN	0.93	0.93	0.93	0.93	0.93	1.0	0.26	NaN
...	...	...	...	...	...	...	...	...	...	...	...	...	...
2018	QUINDIO	2.21	1.0	1.08	NaN	1.08	1.08	1.08	1.08	1.08	1.0	2.07	NaN
	RISARALDA	4.83	1.0	1.28	NaN	1.28	1.28	1.28	1.28	1.28	1.0	5.37	NaN
	SANTANDER	5.69	1.0	1.32	NaN	1.32	1.32	1.32	1.32	1.32	1.0	6.53	NaN
	TOLIMA	13.11	1.0	1.00	NaN	1.00	1.00	1.00	1.00	1.00	1.0	11.39	NaN
	VALLE DEL CAUCA	6.51	1.0	1.03	NaN	1.03	1.03	1.03	1.03	1.03	1.0	5.80	NaN

266 rows × 16 columns



```
In [59]: produccion_Anio=produccion_df.groupby(["Anio"]).describe()
produccion_Anio
```

Out[59]:

	Rendimiento (ha/ton)								Produccion Nacional (ton)				
	count	mean	std	min	25%	50%	75%	max	count	mean	...	75%	m
Anio													
2007	22.0	0.950455	0.279566	0.45	0.7900	0.900	1.1525	1.53	22.0	4.545909	...	7.8475	14
2008	22.0	0.982727	0.322670	0.45	0.7775	0.905	1.2000	1.79	22.0	4.545455	...	7.7600	14
2009	22.0	0.881364	0.264652	0.30	0.7600	0.930	1.1125	1.21	22.0	4.545455	...	7.3425	14
2010	23.0	0.906087	0.324692	0.00	0.7050	0.960	1.1250	1.52	23.0	4.348261	...	7.3550	14
2011	23.0	0.854348	0.238305	0.47	0.6100	0.900	1.0550	1.20	23.0	4.348696	...	7.0800	14
2012	23.0	0.858696	0.329618	0.00	0.7450	0.830	0.9150	2.00	23.0	4.347391	...	6.9850	14
2013	22.0	0.759545	0.145421	0.60	0.6000	0.755	0.8800	0.99	22.0	4.545455	...	6.4400	14
2014	22.0	0.822273	0.157629	0.64	0.6500	0.815	0.9500	1.06	22.0	4.545455	...	6.5950	14
2015	22.0	1.024545	0.110096	0.77	0.9350	1.065	1.1075	1.15	22.0	4.544545	...	6.4675	14
2016	21.0	1.063810	0.116725	0.79	0.9600	1.120	1.1500	1.19	21.0	4.761429	...	6.6800	14
2017	22.0	1.068182	0.272443	0.66	0.8450	1.090	1.2900	1.50	22.0	4.545909	...	6.3550	14
2018	22.0	1.079545	0.296672	0.62	0.8575	1.120	1.3125	1.52	22.0	4.545455	...	6.3475	14

12 rows × 24 columns



```
In [60]: produccion_Anio2_Rendimiento=produccion_df.groupby(["Rendimiento (ha/ton)"]).describe()
produccion_Anio2_Rendimiento
```

Out[60]:

Rendimiento (ha/ton)	Anio									Produccion Nacional		
	count	mean	std	min	25%	50%	75%	max	count	mean	...	...
0.00	2.0	2011.0	1.414214	2010.0	2010.50	2011.0	2011.50	2012.0	2.0	0.000	...	...
0.30	1.0	2009.0	NaN	2009.0	2009.00	2009.0	2009.00	2009.0	1.0	1.440	...	...
0.38	1.0	2009.0	NaN	2009.0	2009.00	2009.0	2009.00	2009.0	1.0	0.040	...	...
0.45	2.0	2007.5	0.707107	2007.0	2007.25	2007.5	2007.75	2008.0	2.0	1.640	...	...
0.47	1.0	2011.0	NaN	2011.0	2011.00	2011.0	2011.00	2011.0	1.0	0.300	...	...
...	...	...	...	...	...	...	...	...	...	...	...	...
1.50	1.0	2017.0	NaN	2017.0	2017.00	2017.0	2017.00	2017.0	1.0	0.600	...	...
1.52	2.0	2014.0	5.656854	2010.0	2012.00	2014.0	2016.00	2018.0	2.0	4.935	...	...
1.53	1.0	2007.0	NaN	2007.0	2007.00	2007.0	2007.00	2007.0	1.0	8.790	...	...
1.79	1.0	2008.0	NaN	2008.0	2008.00	2008.0	2008.00	2008.0	1.0	9.440	...	...
2.00	1.0	2012.0	NaN	2012.0	2012.00	2012.0	2012.00	2012.0	1.0	0.020	...	...

94 rows × 24 columns



```
In [61]: produccion_df.describe()
# Indica datos estadísticos generales del dataframe produccion desde el año 2007
```

Out[61]:

	Anio	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
count	266.000000	266.000000	266.000000	266.000000
mean	2012.469925	0.936429	4.511316	4.511203
std	3.443484	0.267129	4.950568	4.565865
min	2007.000000	0.000000	0.000000	0.000000
25%	2010.000000	0.750000	0.352500	0.390000
50%	2012.000000	0.940000	2.720000	3.120000
75%	2015.000000	1.120000	7.147500	6.875000
max	2018.000000	2.000000	18.670000	16.430000



```
In [62]: produccion_df["Produccion Nacional (ton)"].describe()  
# Indica datos estadísticos generales para la Producción nacional del dataframe p
```

```
Out[62]: count      266.000000  
mean         4.511316  
std          4.950568  
min          0.000000  
25%          0.352500  
50%          2.720000  
75%          7.147500  
max          18.670000  
Name: Produccion Nacional (ton), dtype: float64
```

```
In [63]: produccion_df["Area Nacional (ha)"].describe()  
# Indica datos estadísticos generales para el Area Nacional del dataframe producc
```

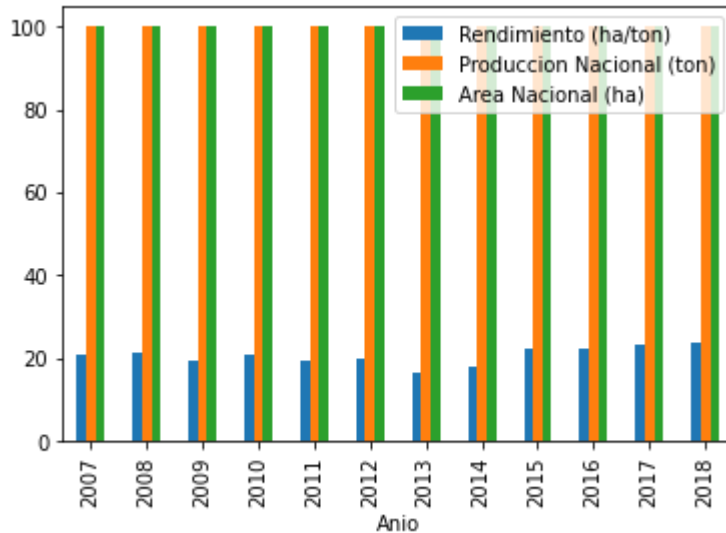
```
Out[63]: count      266.000000  
mean         4.511203  
std          4.565865  
min          0.000000  
25%          0.390000  
50%          3.120000  
75%          6.875000  
max          16.430000  
Name: Area Nacional (ha), dtype: float64
```

```
In [64]: produccion_df.describe()  
produccion_df.mean()  
# Indica el promedio del dataframe produccion para Rendimiento, Produccion y el A
```

```
Out[64]: Anio                2012.469925  
Rendimiento (ha/ton)         0.936429  
Produccion Nacional (ton)    4.511316  
Area Nacional (ha)          4.511203  
dtype: float64
```

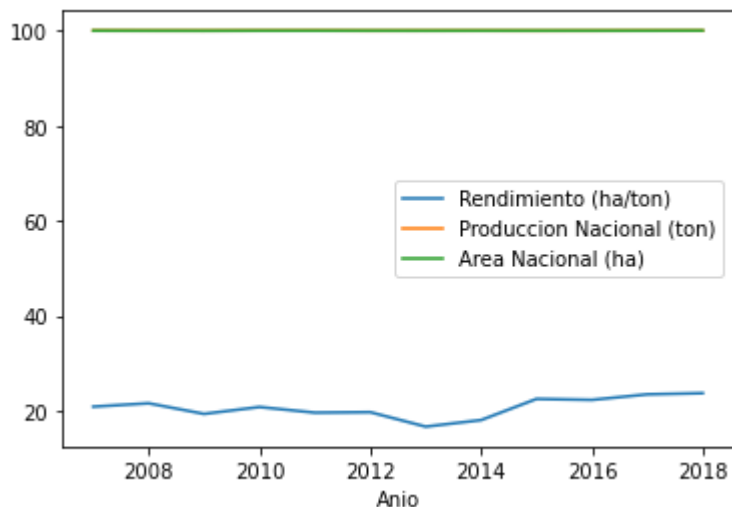
```
In [65]: import numpy as np
import re
import sys
%matplotlib inline
produccion_grouped_Anio2.plot(kind='bar')
```

Out[65]: <matplotlib.axes.\_subplots.AxesSubplot at 0x19158226fa0>



```
In [66]: import numpy as np
import re
import sys
%matplotlib inline
produccion_grouped_Anio2.plot(kind='line')
```

Out[66]: <matplotlib.axes.\_subplots.AxesSubplot at 0x191597935e0>

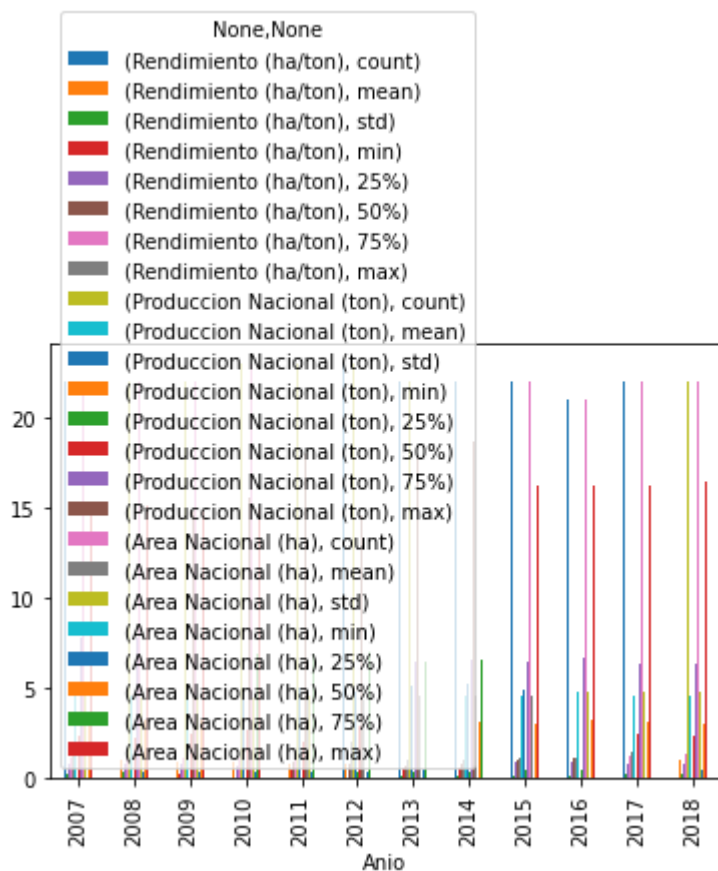


```
In [67]: produccion_df.duplicated().sum()
#Registros que esten duplicados
```

Out[67]: 0

```
In [68]: # Construcción del gráfico produccion por año tipo lineas
import numpy as np
import re
import sys
%matplotlib inline
produccion_Anio.plot(kind='bar')
```

Out[68]: <matplotlib.axes.\_subplots.AxesSubplot at 0x19159817880>



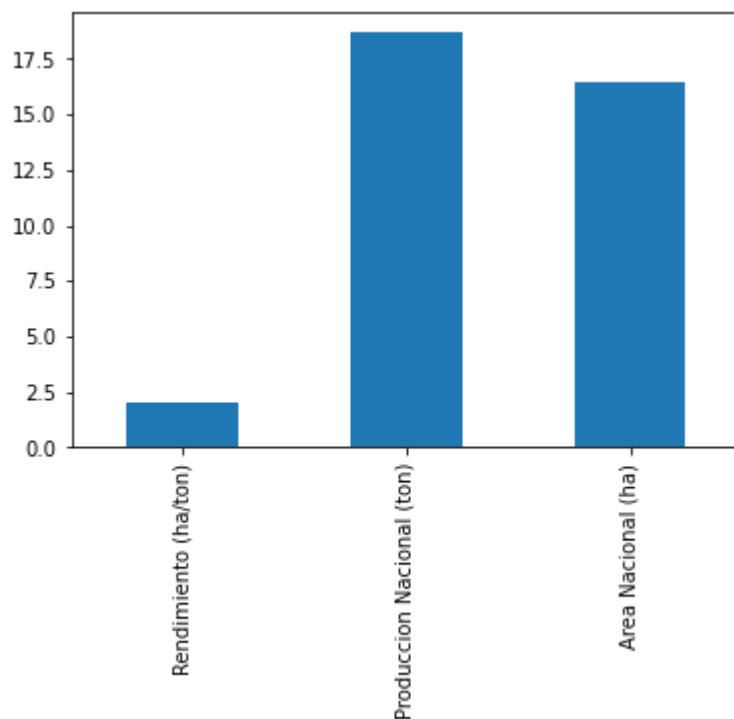
```
In [69]: produccion_grouped_Departamento1=produccion_df.groupby(["Anio", "Departamento"]).s
produccion_grouped_Departamento1
```

Out[69]: Rendimiento (ha/ton) 2.00  
 Produccion Nacional (ton) 18.67  
 Area Nacional (ha) 16.43  
 dtype: float64

In [70]:

```
produccion_grouped_Departamento1.plot(kind='bar')
```

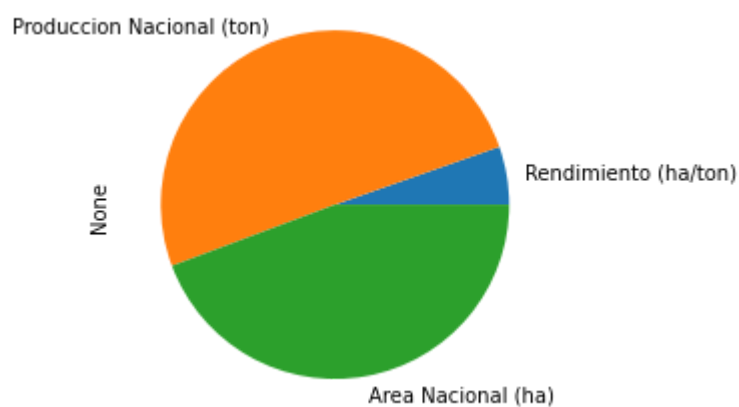
Out[70]: <matplotlib.axes.\_subplots.AxesSubplot at 0x19158340af0>



In [71]:

```
produccion_grouped_Departamento1.plot(kind='pie')
```

Out[71]: <matplotlib.axes.\_subplots.AxesSubplot at 0x19158323310>



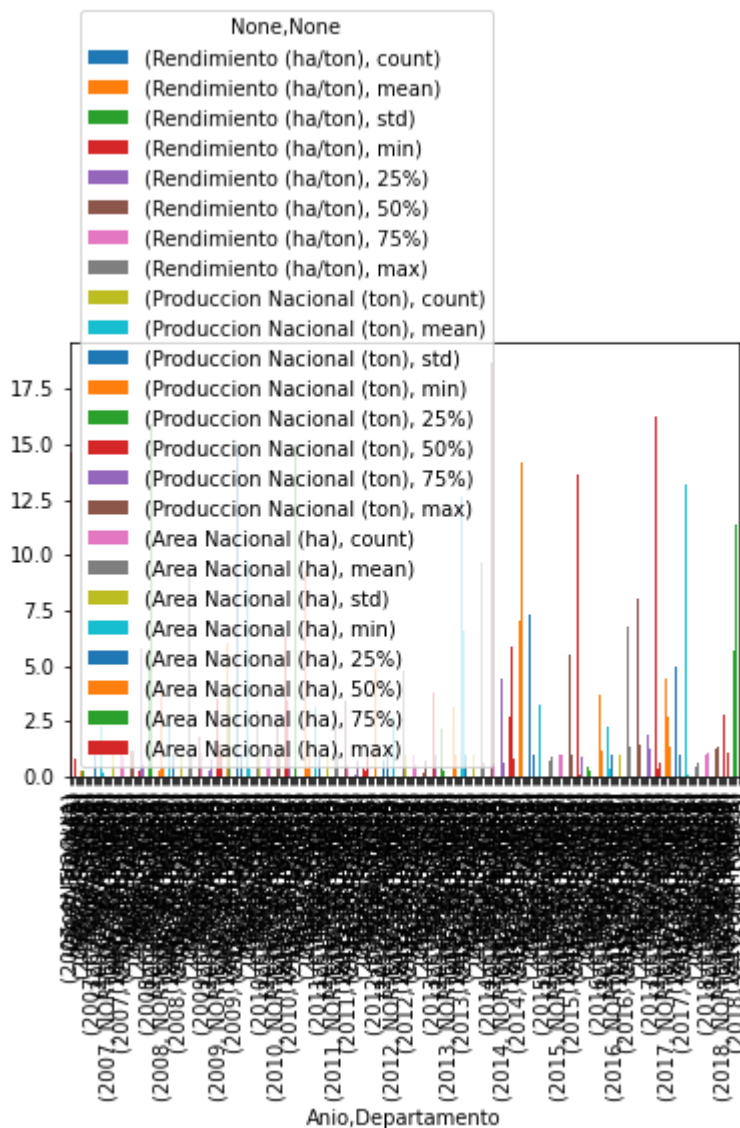
```
In [18]: # Construcción del gráfico Rendimiento por año tipo lineas
%matplotlib inline
produccion_Anio_Rendimiento.plot(kind='bar')
```

```
-----
NameError                                Traceback (most recent call last)
<ipython-input-18-cde4741a05ea> in <module>
      1 # Construcción del gráfico Rendimiento por año tipo lineas
      2 get_ipython().run_line_magic('matplotlib', 'inline')
----> 3 produccion_Anio_Rendimiento.plot(kind='bar')
```

**NameError:** name 'produccion\_Anio\_Rendimiento' is not defined

```
In [73]: # Construcción del gráfico produccion por departamento año tipo lineas
%matplotlib inline
produccion_grouped_Departamento.plot(kind='bar')
```

Out[73]: <matplotlib.axes.\_subplots.AxesSubplot at 0x19158283ee0>



```
In [74]: grouped_data = produccion_df.groupby("Departamento")
z=grouped_data.describe().mean()
print (z)
```

Anio	count	11.083333
	mean	2012.382576
	std	3.479313
	min	2007.333333
	25%	2009.854167
	50%	2012.375000
	75%	2014.895833
	max	2017.458333
Rendimiento (ha/ton)	count	11.083333
	mean	0.889467
	std	0.216119
	min	0.620833
	25%	0.769167
	50%	0.863750
	75%	0.986771
	max	1.235417
Produccion Nacional (ton)	count	11.083333
	mean	4.166733
	std	0.719931
	min	3.261250
	25%	3.687812
	50%	4.031667
	75%	4.614271
	max	5.387500
Area Nacional (ha)	count	11.083333
	mean	4.166632
	std	0.511340
	min	3.537500
	25%	3.758229
	50%	4.136042
	75%	4.588854
	max	4.838333

dtype: float64

```
In [75]: departamentos_counts = produccion_df.groupby("Departamento")["Producto"].count()
print(departamentos_counts)
# Permite verificar y contar para cada uno de los Departamentos las distintas variables
# Se encuentra que algunos departamentos tienen otros valores diferentes a los 12
```

```
Departamento
ANTIOQUIA      12
ARAUCA         2
BOLIVAR        12
BOYACA         12
CALDAS         12
CAQUETA        12
CASANARE       12
CAUCA          12
CESAR          12
CHOCO          12
CUNDINAMARCA   12
GUAVIARE       1
HUILA          12
LA GUAJIRA     12
MAGDALENA      12
META           12
NARIÑO         12
NORTE DE SANTANDER 12
PUTUMAYO       11
QUINDIO        12
RISARALDA      12
SANTANDER      12
TOLIMA         12
VALLE DEL CAUCA 12
Name: Producto, dtype: int64
```

```
In [76]: Grupos_Departamentos=produccion_df.groupby("Año")["Departamento"].count()
print (Grupos_Departamentos)
# Indica la cantidad de Departamentos incluidos o analizados en cada uno de los Años
```

```
Año
2007    22
2008    22
2009    22
2010    23
2011    23
2012    23
2013    22
2014    22
2015    22
2016    21
2017    22
2018    22
Name: Departamento, dtype: int64
```

```
In [77]: Departamento_Meta=produccion_df.loc[produccion_df["Departamento"]=="META"]
print (Departamento_Meta)
# Indica los resultados estadísticos por año para el Departamento Seleccionado
```

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	\
13	2007	META	CAFE	2,048.00	1,617.20	
35	2008	META	CAFE	2,146.00	1,656.96	
57	2009	META	CAFE	2,216.00	1,672.60	
80	2010	META	CAFE	2,326.00	2,221.90	
103	2011	META	CAFE	2,578.00	2,533.75	
126	2012	META	CAFE	2,783.00	2,133.10	
148	2013	META	CAFE	2,483.43	1,650.41	
170	2014	META	CAFE	2,739.71	1,950.84	
192	2015	META	CAFE	2,922.21	3,206.35	
214	2016	META	CAFE	2,924.89	3,322.42	
235	2017	META	CAFE	2,926.85	4,013.11	
257	2018	META	CAFE	2,761.01	3,877.62	

	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
13	0.79	0.20	0.27
35	0.77	0.20	0.28
57	0.75	0.24	0.29
80	0.96	0.29	0.31
103	0.98	0.40	0.36
126	0.77	0.34	0.39
148	0.66	0.25	0.32
170	0.71	0.27	0.34
192	1.10	0.38	0.36
214	1.14	0.39	0.38
235	1.37	0.47	0.39
257	1.40	0.45	0.37



```
In [78]: Departamento_QUINDIO=produccion_df.loc[produccion_df["Departamento"]=="QUINDIO"]
print (Departamento_QUINDIO)
# Indica los resultados estadísticos por año para el Departamento Seleccionado
```

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	\
17	2007	QUINDIO	CAFE	19,904.00	25,426.00	
39	2008	QUINDIO	CAFE	19,571.00	23,669.00	
61	2009	QUINDIO	CAFE	19,052.00	21,985.00	
84	2010	QUINDIO	CAFE	18,159.00	21,065.00	
107	2011	QUINDIO	CAFE	20,139.30	20,814.11	
130	2012	QUINDIO	CAFE	21,109.83	18,030.13	
152	2013	QUINDIO	CAFE	21,203.03	20,599.27	
174	2014	QUINDIO	CAFE	21,462.81	22,518.42	
196	2015	QUINDIO	CAFE	21,491.21	24,694.56	
217	2016	QUINDIO	CAFE	20,041.70	23,791.30	
239	2017	QUINDIO	CAFE	17,699.67	18,792.05	
261	2018	QUINDIO	CAFE	16,374.73	17,739.03	

	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
17	1.28	3.07	2.60
39	1.21	2.86	2.58
61	1.15	3.10	2.52
84	1.16	2.70	2.44
107	1.03	3.25	2.83
130	0.85	2.88	2.97
152	0.97	3.16	2.75
174	1.05	3.09	2.70
196	1.15	2.90	2.68
217	1.19	2.79	2.58
239	1.06	2.21	2.35
261	1.08	2.07	2.21

```
In [79]: Estadística_Anio2015=produccion_df.loc[produccion_df["Anio"]== 2015]
print (Estadística_Anio2015)
# Indica los resultados estadísticos por departamento para el año 2015
```

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	\
179	2015	ANTIOQUIA	CAFE	109,649.61	120,365.77	
180	2015	BOLIVAR	CAFE	1,065.07	1,089.74	
181	2015	BOYACA	CAFE	10,461.85	9,501.54	
182	2015	CALDAS	CAFE	58,376.40	67,231.37	
183	2015	CAQUETA	CAFE	3,410.56	3,749.27	
184	2015	CASANARE	CAFE	2,752.31	2,626.73	
185	2015	CAUCA	CAFE	77,405.83	83,626.44	
186	2015	CESAR	CAFE	25,948.50	22,240.81	
187	2015	CHOCO	CAFE	137.47	158.20	
188	2015	CUNDINAMARCA	CAFE	34,101.49	31,165.15	
189	2015	HUILA	CAFE	130,452.40	145,168.10	
190	2015	LA GUAJIRA	CAFE	5,631.53	4,317.50	
191	2015	MAGDALENA	CAFE	17,996.31	16,691.31	
192	2015	META	CAFE	2,922.21	3,206.35	
193	2015	NARIÑO	CAFE	33,490.93	36,607.56	
194	2015	NORTE DE SANTANDER	CAFE	22,940.64	20,267.64	
195	2015	PUTUMAYO	CAFE	128.65	124.67	
196	2015	QUINDIO	CAFE	21,491.21	24,694.56	
197	2015	RISARALDA	CAFE	41,732.03	47,215.69	
198	2015	SANTANDER	CAFE	42,679.11	47,304.16	
199	2015	TOLIMA	CAFE	103,368.73	105,563.88	
200	2015	VALLE DEL CAUCA	CAFE	54,938.79	57,583.56	

	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
179	1.10	14.15	13.69
180	1.02	0.13	0.13
181	0.91	1.12	1.31
182	1.15	7.90	7.29
183	1.10	0.44	0.43
184	0.95	0.31	0.34
185	1.08	9.83	9.66
186	0.86	2.62	3.24
187	1.15	0.02	0.02
188	0.91	3.66	4.26
189	1.11	17.07	16.28
190	0.77	0.51	0.70
191	0.93	1.96	2.25
192	1.10	0.38	0.36
193	1.09	4.30	4.18
194	0.88	2.38	2.86
195	0.97	0.01	0.02
196	1.15	2.90	2.68
197	1.13	5.55	5.21
198	1.11	5.56	5.33
199	1.02	12.41	12.90
200	1.05	6.77	6.86

```
In [80]: Estadística_Anio2018=produccion_df.loc[produccion_df["Anio"]== 2018]
print (Estadística_Anio2018)
# Indica los resultados estadísticos por departamento para el año 2018
```

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	\
244	2018	ANTIOQUIA	CAFE	98,038.15	141,898.91	
245	2018	BOLIVAR	CAFE	1,182.13	734.91	
246	2018	BOYACA	CAFE	9,653.45	7,780.34	
247	2018	CALDAS	CAFE	50,762.22	68,670.96	
248	2018	CAQUETA	CAFE	3,485.24	5,280.40	
249	2018	CASANARE	CAFE	2,360.55	1,629.25	
250	2018	CAUCA	CAFE	82,085.54	102,147.00	
251	2018	CESAR	CAFE	23,915.45	14,943.62	
252	2018	CHOCO	CAFE	140.33	181.42	
253	2018	CUNDINAMARCA	CAFE	29,085.24	32,580.24	
254	2018	HUILA	CAFE	122,002.46	136,161.86	
255	2018	LA GUAJIRA	CAFE	4,810.97	2,990.91	
256	2018	MAGDALENA	CAFE	17,414.32	10,826.24	
257	2018	META	CAFE	2,761.01	3,877.62	
258	2018	NARIÑO	CAFE	33,465.54	35,679.42	
259	2018	NORTE DE SANTANDER	CAFE	20,873.04	23,471.69	
260	2018	PUTUMAYO	CAFE	209.93	289.50	
261	2018	QUINDIO	CAFE	16,374.73	17,739.03	
262	2018	RISARALDA	CAFE	35,874.73	45,918.75	
263	2018	SANTANDER	CAFE	42,269.07	55,918.71	
264	2018	TOLIMA	CAFE	97,304.04	97,451.31	
265	2018	VALLE DEL CAUCA	CAFE	48,305.31	49,667.88	

	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
244	1.45	16.58	13.21
245	0.62	0.09	0.16
246	0.81	0.91	1.30
247	1.35	8.02	6.84
248	1.52	0.62	0.47
249	0.69	0.19	0.32
250	1.24	11.94	11.06
251	0.62	1.75	3.22
252	1.29	0.02	0.02
253	1.12	3.81	3.92
254	1.12	15.91	16.43
255	0.62	0.35	0.65
256	0.62	1.26	2.35
257	1.40	0.45	0.37
258	1.07	4.17	4.51
259	1.12	2.74	2.81
260	1.38	0.03	0.03
261	1.08	2.07	2.21
262	1.28	5.37	4.83
263	1.32	6.53	5.69
264	1.00	11.39	13.11
265	1.03	5.80	6.51

```
In [81]: produccion_df[0:10]
#Lista los primeros 10 elementos del dataframe
```

Out[81]:

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
0	2007	ANTIOQUIA	CAFE	112,343.60	120,500.80	1.07	14.54	14.66
1	2007	BOLIVAR	CAFE	502.00	446.00	0.89	0.05	0.07
2	2007	BOYACA	CAFE	11,374.50	9,683.10	0.85	1.17	1.48
3	2007	CALDAS	CAFE	78,393.65	92,815.00	1.18	11.20	10.23
4	2007	CAQUETA	CAFE	2,295.00	2,134.00	0.93	0.26	0.30
5	2007	CASANARE	CAFE	2,605.00	2,048.40	0.79	0.25	0.34
6	2007	CAUCA	CAFE	53,471.00	51,348.00	0.96	6.19	6.98
7	2007	CESAR	CAFE	23,172.00	13,278.50	0.57	1.60	3.02
8	2007	CHOCO	CAFE	290.00	205.90	0.71	0.02	0.04
9	2007	CUNDINAMARCA	CAFE	43,017.30	33,729.14	0.78	4.07	5.61

```
In [82]: produccion_df[11:30]
#Lista los elementos desde el 11 al 30 del dataframe
```

Out[82]:

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
11	2007	LA GUAJIRA	CAFE	4,785.00	2,958.70	0.62	0.36	0.62
12	2007	MAGDALENA	CAFE	17,506.00	14,005.00	0.80	1.69	2.28
13	2007	META	CAFE	2,048.00	1,617.20	0.79	0.20	0.27
14	2007	NARIÑO	CAFE	24,458.50	31,770.05	1.30	3.83	3.19
15	2007	NORTE DE SANTANDER	CAFE	30,171.84	13,593.24	0.45	1.64	3.94
16	2007	PUTUMAYO	CAFE	35.00	34.00	0.97	0.00	0.00
17	2007	QUINDIO	CAFE	19,904.00	25,426.00	1.28	3.07	2.60
18	2007	RISARALDA	CAFE	47,689.25	72,842.55	1.53	8.79	6.22
19	2007	SANTANDER	CAFE	34,406.67	29,469.52	0.86	3.56	4.49
20	2007	TOLIMA	CAFE	91,679.10	112,322.38	1.23	13.55	11.96
21	2007	VALLE DEL CAUCA	CAFE	76,667.80	69,618.24	0.91	8.40	10.00
22	2008	ANTIOQUIA	CAFE	114,694.00	113,505.20	0.99	13.70	15.13
23	2008	BOLIVAR	CAFE	572.00	711.00	1.24	0.09	0.08
24	2008	BOYACA	CAFE	10,778.50	9,547.30	0.89	1.15	1.42
25	2008	CALDAS	CAFE	74,897.00	86,884.00	1.16	10.49	9.88
26	2008	CAQUETA	CAFE	2,735.00	2,469.00	0.90	0.30	0.36
27	2008	CASANARE	CAFE	2,149.00	1,388.13	0.65	0.17	0.28
28	2008	CAUCA	CAFE	56,208.00	48,073.00	0.86	5.80	7.41
29	2008	CESAR	CAFE	23,198.00	13,841.45	0.60	1.67	3.06

```
In [83]: # Recuerda que en INTERNET hay MUCHAS más instrucciones que puedes APLICAR en tu
# Solamente te he mostrado algunas instrucciones importantes, pero hay muchísimas
```

```
In [84]: # ANIMO, ESFUERZATE. SOLO TU MISMO TE IMPONES LOS LIMITES, DEBES VENCER TODOS LOS
# INSTRUCTOR : Ing. Luis Armando Amaya Quiroga
```

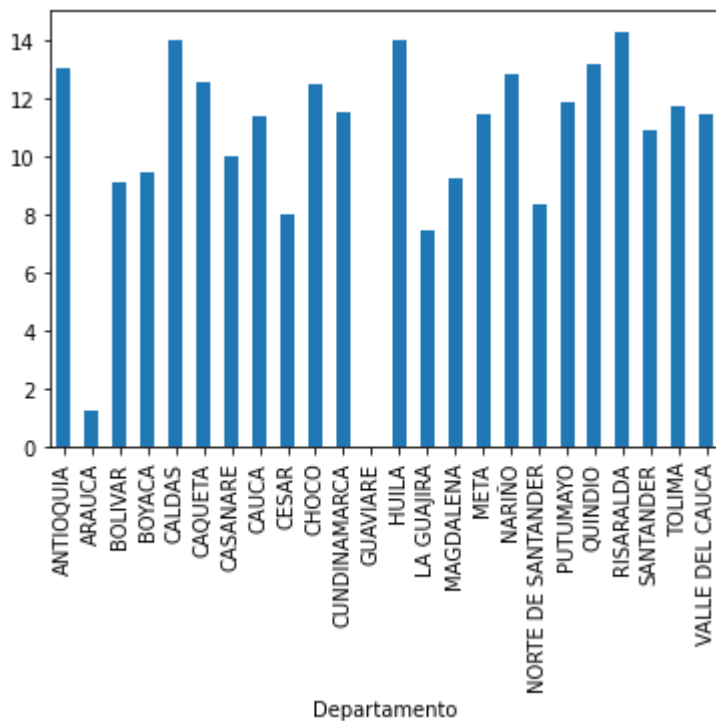
```
In [85]: Grupos_Departamentos_Rendimiento=produccion_df.groupby("Departamento")["Rendimier  
Grupos_Departamentos_Rendimiento  
# Indica el Rendimiento total del café en hectareas por toneladas (ha/ton) de cac
```

```
Out[85]: Departamento  
ANTIOQUIA      13.01  
ARAUCA         1.20  
BOLIVAR        9.07  
BOYACA         9.41  
CALDAS        14.00  
CAQUETA       12.54  
CASANARE       10.01  
CAUCA         11.36  
CESAR          7.95  
CHOCO         12.50  
CUNDINAMARCA  11.48  
GUAVIARE       0.00  
HUILA         13.98  
LA GUAJIRA     7.45  
MAGDALENA     9.21  
META          11.40  
NARIÑO        12.79  
NORTE DE SANTANDER  8.36  
PUTUMAYO      11.84  
QUINDIO       13.18  
RISARALDA     14.29  
SANTANDER     10.88  
TOLIMA        11.73  
VALLE DEL CAUCA 11.45  
Name: Rendimiento (ha/ton), dtype: float64
```

In [86]:

```
Grupos_Departamentos_Rendimiento.plot(kind='bar')
```

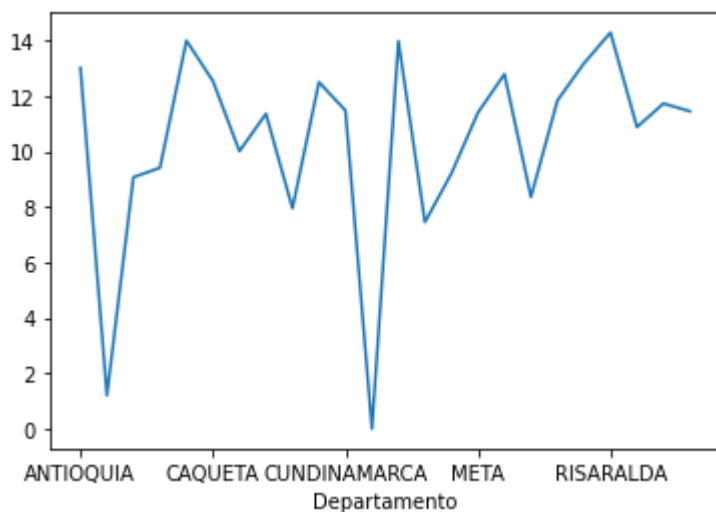
Out[86]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1915d1ab250>



In [87]:

```
Grupos_Departamentos_Rendimiento.plot(kind='line')
```

Out[87]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1915c7f0730>



```
In [88]: Grupos_Departamentos_Produccion=produccion_df.groupby("Departamento")["Produccion"]
Grupos_Departamentos_Produccion
# Indica la Produccion total del café en hectareas por toneladas (ha/ton) de cada
```

```
Out[88]: Departamento
ANTIOQUIA      120,500.80113,505.20103,703.00121,253.38115,26...
ARAUCA                                0.0012.00
BOLIVAR        446.00711.00292.60510.00510.00652.50395.07606....
BOYACA          9,683.109,547.308,567.977,083.075,643.394,981....
CALDAS          92,815.0086,884.0081,668.2295,957.9078,805.875...
CAQUETA         2,134.002,469.002,332.002,902.502,528.402,446....
CASANARE        2,048.401,388.132,079.702,564.862,023.501,718....
CAUCA           51,348.0048,073.0047,221.0045,113.0041,645.395...
CESAR            13,278.5013,841.4512,770.0013,276.0811,035.851...
CHOCO            205.9068.0078.7598.0098.00140.00105.93125.4215...
CUNDINAMARCA    33,729.1478,254.7737,118.0737,214.8032,780.353...
GUAVIARE                                0.00
HUILA           129,052.51131,316.47104,609.42104,336.5685,150...
LA GUAJIRA      2,958.702,328.902,340.402,393.001,933.003,434....
MAGDALENA       14,005.0014,017.0013,412.8013,600.0013,301.601...
META            1,617.201,656.961,672.602,221.902,533.752,133....
NARIÑO          31,770.0531,262.5027,487.7124,594.1024,073.952...
NORTE DE SANTANDER 13,593.2413,593.2510,221.6922,111.6512,332.001...
PUTUMAYO        34.0035.6026.7026.7045.8048.4016.8776.04124.67...
QUINDIO         25,426.0023,669.0021,985.0021,065.0020,814.111...
RISARALDA       72,842.5560,079.0053,648.0072,091.0049,042.313...
SANTANDER       29,469.5229,016.7526,311.6127,094.1622,089.822...
TOLIMA          112,322.38101,201.8888,633.1094,230.2053,288.4...
VALLE DEL CAUCA 69,618.2465,666.4362,711.0869,496.6565,475.636...
Name: Produccion (ton), dtype: object
```

```
In [89]: produccion_df["Rendimiento (ha/ton)"].describe()
# Indica datos estadísticos generales para el Rendimiento del dataframe produccion
```

```
Out[89]: count      266.000000
mean         0.936429
std          0.267129
min          0.000000
25%          0.750000
50%          0.940000
75%          1.120000
max          2.000000
Name: Rendimiento (ha/ton), dtype: float64
```

```
In [90]: produccion_df["Produccion (ton)"].describe()
# Indica datos estadísticos generales para la Produccion del dataframe produccion
```

```
Out[90]: count      266
unique      262
top         98.00
freq        2
Name: Produccion (ton), dtype: object
```



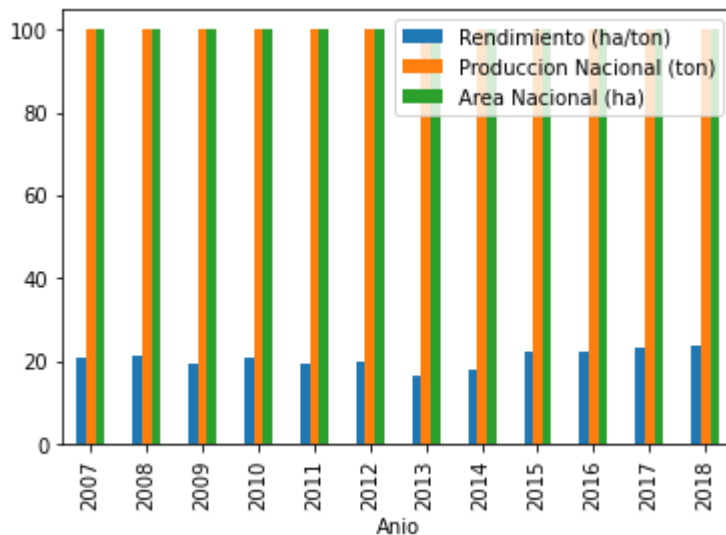
```
In [91]: produccion_grouped_Anio4=produccion_df.groupby("Anio").sum()
produccion_grouped_Anio4
```

Out[91]:

	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
Anio			
2007	20.91	100.01	100.00
2008	21.62	100.00	99.99
2009	19.39	100.00	99.98
2010	20.84	100.01	100.00
2011	19.65	100.02	100.00
2012	19.75	99.99	100.00
2013	16.71	100.00	99.99
2014	18.09	100.00	100.00
2015	22.54	99.98	100.00
2016	22.34	99.99	100.00
2017	23.50	100.01	100.00
2018	23.75	100.00	100.02

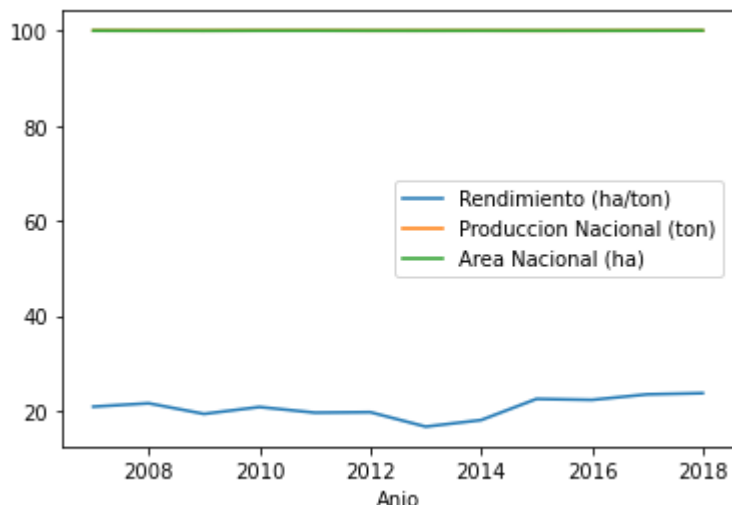
```
In [92]: import numpy as np
import re
import sys
%matplotlib inline
produccion_grouped_Anio4.plot(kind='bar')
```

Out[92]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1915cf67610>



```
In [93]: import numpy as np
import re
import sys
%matplotlib inline
produccion_grouped_Anio4.plot(kind='line')
```

Out[93]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1915ec63070>



```
In [94]: produccion_grouped_Anio5=produccion_df.groupby("Anio")["Departamento"].sum()
produccion_grouped_Anio5
```

Out[94]: Anio

2007	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2008	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2009	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2010	ANTIOQUIAARAUCABOLIVARBOYACACALDASCAQUETACASAN...
2011	ANTIOQUIAARAUCABOLIVARBOYACACALDASCAQUETACASAN...
2012	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2013	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2014	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2015	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2016	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2017	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...
2018	ANTIOQUIABOLIVARBOYACACALDASCAQUETACASANARECAU...

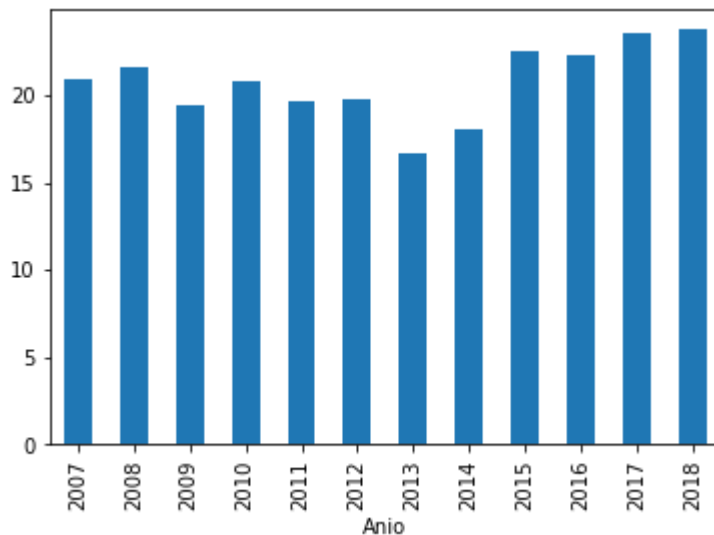
Name: Departamento, dtype: object

```
In [95]: Grupos_Departamentos_Rendimiento4=produccion_df.groupby("Anio")["Rendimiento (ha/ton)"]  
Grupos_Departamentos_Rendimiento4  
# Indica el Rendimiento total del café en hectareas por toneladas (ha/ton) por año
```

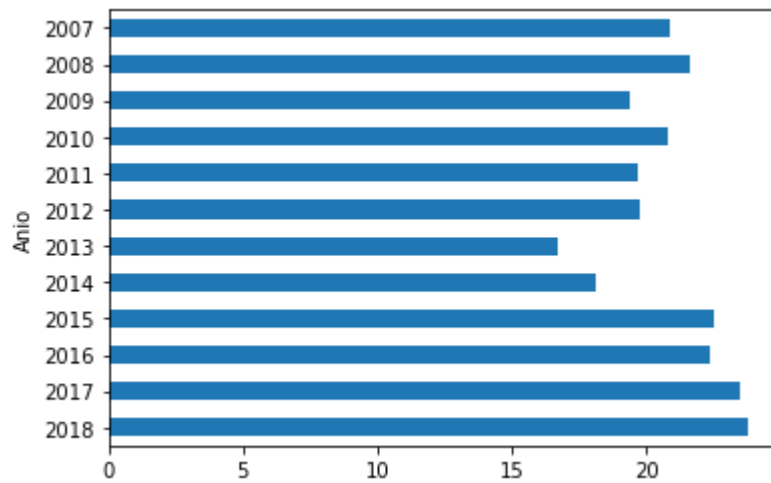
```
Out[95]: Anio  
2007      20.91  
2008      21.62  
2009      19.39  
2010      20.84  
2011      19.65  
2012      19.75  
2013      16.71  
2014      18.09  
2015      22.54  
2016      22.34  
2017      23.50  
2018      23.75  
Name: Rendimiento (ha/ton), dtype: float64
```

```
In [96]: import numpy as np  
import re  
import sys  
%matplotlib inline  
Grupos_Departamentos_Rendimiento4.plot(kind='bar')
```

```
Out[96]: <matplotlib.axes._subplots.AxesSubplot at 0x1915f23b6d0>
```



```
In [97]: #pd.value_counts(produccion_df['Departamento']).plot(kind='barh').invert_yaxis()  
Grupos_Departamentos_Rendimiento4.plot(kind='barh').invert_yaxis()
```



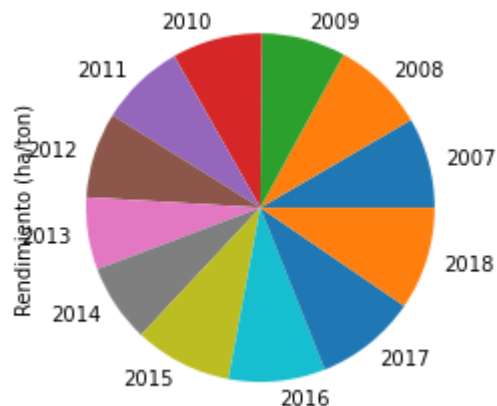
```
In [98]: import numpy as np  
import re  
import sys  
%matplotlib inline  
Grupos_Departamentos_Rendimiento4.plot(kind='line')
```

Out[98]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1915f49f0a0>



```
In [99]: import numpy as np
import re
import sys
%matplotlib inline
Grupos_Departamentos_Rendimiento4.plot(kind='pie')
```

Out[99]: <matplotlib.axes.\_subplots.AxesSubplot at 0x1915f272e20>



```
In [100]: produccion_df.isnull()
#verificar si existen valores nulos
```

Out[100]:

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
0	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False
...	...	...	...	...	...	...	...	...
261	False	False	False	False	False	False	False	False
262	False	False	False	False	False	False	False	False
263	False	False	False	False	False	False	False	False
264	False	False	False	False	False	False	False	False
265	False	False	False	False	False	False	False	False

266 rows × 8 columns

```
In [101]: produccion_df.isnull().sum()
## Totales de los valores nulos en el dataframe
```

```
Out[101]: Anio                                0
Departamento                             0
Producto                                  0
Area (ha)                                0
Produccion (ton)                          0
Rendimiento (ha/ton)                      0
Produccion Nacional (ton)                 0
Area Nacional (ha)                       0
dtype: int64
```

```
describe(produccion_df[,6:18])
```

```
In [102]: produccion_df.tail()
```

```
Out[102]:
```

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
<b>261</b>	2018	QUINDIO	CAFE	16,374.73	17,739.03	1.08	2.07	2.21
<b>262</b>	2018	RISARALDA	CAFE	35,874.73	45,918.75	1.28	5.37	4.83
<b>263</b>	2018	SANTANDER	CAFE	42,269.07	55,918.71	1.32	6.53	5.69
<b>264</b>	2018	TOLIMA	CAFE	97,304.04	97,451.31	1.00	11.39	13.11
<b>265</b>	2018	VALLE DEL CAUCA	CAFE	48,305.31	49,667.88	1.03	5.80	6.51

```
In [103]: produccion_df.describe()
```

```
Out[103]:
```

	Anio	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
<b>count</b>	266.000000	266.000000	266.000000	266.000000
<b>mean</b>	2012.469925	0.936429	4.511316	4.511203
<b>std</b>	3.443484	0.267129	4.950568	4.565865
<b>min</b>	2007.000000	0.000000	0.000000	0.000000
<b>25%</b>	2010.000000	0.750000	0.352500	0.390000
<b>50%</b>	2012.000000	0.940000	2.720000	3.120000
<b>75%</b>	2015.000000	1.120000	7.147500	6.875000
<b>max</b>	2018.000000	2.000000	18.670000	16.430000

In [104]: `produccion_df`

Out[104]:

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
<b>0</b>	2007	ANTIOQUIA	CAFE	112,343.60	120,500.80	1.07	14.54	14.66
<b>1</b>	2007	BOLIVAR	CAFE	502.00	446.00	0.89	0.05	0.07
<b>2</b>	2007	BOYACA	CAFE	11,374.50	9,683.10	0.85	1.17	1.48
<b>3</b>	2007	CALDAS	CAFE	78,393.65	92,815.00	1.18	11.20	10.23
<b>4</b>	2007	CAQUETA	CAFE	2,295.00	2,134.00	0.93	0.26	0.30
...	...	...	...	...	...	...	...	...
<b>261</b>	2018	QUINDIO	CAFE	16,374.73	17,739.03	1.08	2.07	2.21
<b>262</b>	2018	RISARALDA	CAFE	35,874.73	45,918.75	1.28	5.37	4.83
<b>263</b>	2018	SANTANDER	CAFE	42,269.07	55,918.71	1.32	6.53	5.69
<b>264</b>	2018	TOLIMA	CAFE	97,304.04	97,451.31	1.00	11.39	13.11
<b>265</b>	2018	VALLE DEL CAUCA	CAFE	48,305.31	49,667.88	1.03	5.80	6.51

266 rows × 8 columns

In [ ]:

In [ ]: `#pip install pandas-profiling`

In [106]: produccion\_df

Out[106]:

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
<b>0</b>	2007	ANTIOQUIA	CAFE	112,343.60	120,500.80	1.07	14.54	14.66
<b>1</b>	2007	BOLIVAR	CAFE	502.00	446.00	0.89	0.05	0.07
<b>2</b>	2007	BOYACA	CAFE	11,374.50	9,683.10	0.85	1.17	1.48
<b>3</b>	2007	CALDAS	CAFE	78,393.65	92,815.00	1.18	11.20	10.23
<b>4</b>	2007	CAQUETA	CAFE	2,295.00	2,134.00	0.93	0.26	0.30
...	...	...	...	...	...	...	...	...
<b>261</b>	2018	QUINDIO	CAFE	16,374.73	17,739.03	1.08	2.07	2.21
<b>262</b>	2018	RISARALDA	CAFE	35,874.73	45,918.75	1.28	5.37	4.83
<b>263</b>	2018	SANTANDER	CAFE	42,269.07	55,918.71	1.32	6.53	5.69
<b>264</b>	2018	TOLIMA	CAFE	97,304.04	97,451.31	1.00	11.39	13.11
<b>265</b>	2018	VALLE DEL CAUCA	CAFE	48,305.31	49,667.88	1.03	5.80	6.51

266 rows × 8 columns



In [16]: produccion\_df

Out[16]:

	Anio	Departamento	Producto	Area (ha)	Produccion (ton)	Rendimiento (ha/ton)	Produccion Nacional (ton)	Area Nacional (ha)
<b>0</b>	2007	ANTIOQUIA	CAFE	112,343.60	120,500.80	1.07	14.54	14.66
<b>1</b>	2007	BOLIVAR	CAFE	502.00	446.00	0.89	0.05	0.07
<b>2</b>	2007	BOYACA	CAFE	11,374.50	9,683.10	0.85	1.17	1.48
<b>3</b>	2007	CALDAS	CAFE	78,393.65	92,815.00	1.18	11.20	10.23
<b>4</b>	2007	CAQUETA	CAFE	2,295.00	2,134.00	0.93	0.26	0.30
...	...	...	...	...	...	...	...	...
<b>261</b>	2018	QUINDIO	CAFE	16,374.73	17,739.03	1.08	2.07	2.21
<b>262</b>	2018	RISARALDA	CAFE	35,874.73	45,918.75	1.28	5.37	4.83
<b>263</b>	2018	SANTANDER	CAFE	42,269.07	55,918.71	1.32	6.53	5.69
<b>264</b>	2018	TOLIMA	CAFE	97,304.04	97,451.31	1.00	11.39	13.11
<b>265</b>	2018	VALLE DEL CAUCA	CAFE	48,305.31	49,667.88	1.03	5.80	6.51

266 rows × 8 columns

```
In [29]: # USO DE PANDAS PROFILING
# Instructor Ing. Luis Armando Amaya Q.
import pandas as pd
import numpy as np
from pandas_profiling import ProfileReport
profile=ProfileReport(produccion_df, title='CAFE', html={'style': {'full_width':
profile
#NOTA IMPORTANTE
# LA DOS SIGUIENTES INSTRUCCIONES, CREAM UN INFORME EN FORMATO HTML
# DEBE BUSCARLO EN SU COMPUTADOR CON EL NOMBRE:---> ANALISIS EXPLORATORIO CADE_PA
# LUEGO DE ENCONTRAR LA CARPETA ---> Producción_Cafe <-----
# PARA ABRIR EL INFORME DEBE HACER CLIC SOBRE EL ARCHIVO LLAMADO----->your
# RECUERDE: -----> LA DOS SIGUIENTES INSTRUCCIONES, CREAM UN INFORME EN FORMATO H
# TAMBIÉN LE SUBÍ TODA LA CARPETA AL DRIVE CON TODOS ESTOS INFORMES, LLAMADA ----
profile2=profile
profile2.to_file("ANALISIS EXPLORATORIO CAFE_PANDAS.html")
```

Summarize dataset: 100%

22/22 [00:37<00:00, 1.69s/it, Completed]

Generate report structure: 100%

1/1 [00:03<00:00, 3.37s/it]

Render HTML: 100%

1/1 [00:00<00:00, 1.39it/s]

Export report to file: 100%

1/1 [00:00<00:00, 22.79it/s]

```
In [30]: # La siguiente instrucción despliega el informe aquí en JUPYTER NOTEBOOK profile
```

# Overview

## Dataset statistics

<b>Number of variables</b>	8
<b>Number of observations</b>	266
<b>Missing cells</b>	0
<b>Missing cells (%)</b>	0.0%
<b>Duplicate rows</b>	0
<b>Duplicate rows (%)</b>	0.0%
<b>Total size in memory</b>	16.8 KiB
<b>Average record size in memory</b>	64.5 B

## Variable types

<b>NUM</b>	4
<b>CAT</b>	4

## Warnings

Producto has constant value "266"	Constant
Area (ha) has a high cardinality: 261 distinct values	High cardinality
Produccion (ton) has a high cardinality: 262 distinct values	High cardinality
Area Nacional (ha) is highly correlated with Produccion	High correlation

Out[30]:

In [ ]: