

# Data Analytics

January 26, 2021

```
In [2]: import numpy as np
```

```
        ary1 = np.array([1,2,3])
```

```
        print(ary1)
```

```
[1 2 3]
```

```
In [3]: print(type(ary1))
```

```
<class 'numpy.ndarray'>
```

```
In [4]: print(ary1.shape)
```

```
(3,)
```

```
In [5]: ary1[2]
```

```
Out[5]: 3
```

```
In [7]: ary1[2]=5  
        ary1
```

```
Out[7]: array([1, 2, 5])
```

```
In [9]: ary2 = np.array([[1,2,3],[3,4,5]])  
        ary2
```

```
Out[9]: array([[1, 2, 3],  
               [3, 4, 5]])
```

```
In [10]: ary2[1][2]
```

```
Out[10]: 5
```

```
In [11]: print(ary2.shape)
```

(2, 3)

```
In [12]: ary2[1][-1]
```

```
Out[12]: 5
```

```
In [13]: ary2[1][-3]
```

```
Out[13]: 3
```

```
In [14]: arys = np.array(['Chaina', 'India', 'Canada', 'USA'])
        arys
```

```
Out[14]: array(['Chaina', 'India', 'Canada', 'USA'], dtype='<U6')
```

```
In [15]: aryR = np.arange(0,20,2)
```

```
In [87]: aryR = np.arange(0,30,3)
        aryR
```

```
Out[87]: array([ 0,  3,  6,  9, 12, 15, 18, 21, 24, 27])
```

```
In [18]: aryL = np.linspace(0,10,20)
        aryL
```

```
Out[18]: array([ 0.          ,  0.52631579,  1.05263158,  1.57894737,  2.10526316,
                2.63157895,  3.15789474,  3.68421053,  4.21052632,  4.73684211,
                5.26315789,  5.78947368,  6.31578947,  6.84210526,  7.36842105,
                7.89473684,  8.42105263,  8.94736842,  9.47368421, 10.          ])
```

```
In [19]: ary = np.random.rand(10)
        ary
```

```
Out[19]: array([0.56750196, 0.60980437, 0.16004783, 0.69149488, 0.29569082,
                0.62240006, 0.21497158, 0.12443889, 0.22541522, 0.47230696])
```

```
In [20]: ary = np.random.rand(3,4)
        ary
```

```
Out[20]: array([[0.58403665, 0.9600567 , 0.97892197, 0.54657973],
                [0.3440654 , 0.18368897, 0.3425167 , 0.39620669],
                [0.44122453, 0.54291041, 0.69028765, 0.06868475]])
```

```
In [21]: print(np.zeros(10))
        print('/n')
        print(np.zeros((2,3)))
```

```
[0. 0. 0. 0. 0. 0. 0. 0. 0. 0.]
/n
[[0. 0. 0.]
 [0. 0. 0.]]
```

```
In [22]: print(np.ones((10,10,3)))
```

```
[[[1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]
```

```
[[[1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]
```

```
[[[1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]
```

```
[[[1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]  
  [1.  1.  1.]
```

```
[[[1.  1.  1.]  
  [1.  1.  1.]
```

[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]

[[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]]

[[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]]

[[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]]

[[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]  
[1. 1. 1.]

```
[1. 1. 1.]
[1. 1. 1.]
[1. 1. 1.]
[1. 1. 1.]]
```

```
[[1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]
 [1. 1. 1.]]]
```

```
In [23]: ary = [0,1,2]
         print(np.repeat(ary, 5))
```

```
[0 0 0 0 0 1 1 1 1 1 2 2 2 2 2]
```

```
In [24]: print(np.tile(ary,5))
```

```
[0 1 2 0 1 2 0 1 2 0 1 2 0 1 2]
```

```
In [25]: identity_matrix = np.eye(3)
         identity_matrix
```

```
Out[25]: array([[1., 0., 0.],
                [0., 1., 0.],
                [0., 0., 1.]])
```

```
In [26]: np.diag([1,2,3,4,5])
```

```
Out[26]: array([[1, 0, 0, 0, 0],
                [0, 2, 0, 0, 0],
                [0, 0, 3, 0, 0],
                [0, 0, 0, 4, 0],
                [0, 0, 0, 0, 5]])
```

```
In [27]: ary = np.random.rand(5,5)
         ary
```

```
Out[27]: array([[0.25226077, 0.2579296 , 0.46973991, 0.87427238, 0.44335414],
                [0.18210193, 0.41776999, 0.87948047, 0.36747753, 0.07804553],
                [0.39424446, 0.03643247, 0.83427572, 0.39335018, 0.60132669],
                [0.59417161, 0.0712503 , 0.98424964, 0.20258022, 0.33971214],
                [0.6561921 , 0.87046826, 0.86435258, 0.74885978, 0.60192395]])
```

```

In [28]: np.diag(ary)

Out[28]: array([0.25226077, 0.41776999, 0.83427572, 0.20258022, 0.60192395])

In [29]: ary.ndim

Out[29]: 2

In [30]: ary.size

Out[30]: 25

In [31]: ary.shape

Out[31]: (5, 5)

In [32]: np.random.randint(-10,10,5)

Out[32]: array([-1,  1, -5, -2,  6])

In [34]: ary/4

Out[34]: array([[0.06306519, 0.0644824 , 0.11743498, 0.2185681 , 0.11083854],
                [0.04552548, 0.1044425 , 0.21987012, 0.09186938, 0.01951138],
                [0.09856111, 0.00910812, 0.20856893, 0.09833755, 0.15033167],
                [0.1485429 , 0.01781257, 0.24606241, 0.05064505, 0.08492804],
                [0.16404802, 0.21761706, 0.21608814, 0.18721495, 0.15048099]])

In [35]: np.exp(ary)

Out[35]: array([[1.28693158, 1.2942477 , 1.59957811, 2.39713046, 1.55792396],
                [1.19973648, 1.51857135, 2.40964749, 1.44408735, 1.08117189],
                [1.4832631 , 1.03710426, 2.30314532, 1.48193725, 1.82453779],
                [1.81152968, 1.07384997, 2.67580332, 1.22455831, 1.40454323],
                [1.92743884, 2.3880288 , 2.37346895, 2.11458755, 1.82562784]])

In [36]: ary

Out[36]: array([[0.25226077, 0.2579296 , 0.46973991, 0.87427238, 0.44335414],
                [0.18210193, 0.41776999, 0.87948047, 0.36747753, 0.07804553],
                [0.39424446, 0.03643247, 0.83427572, 0.39335018, 0.60132669],
                [0.59417161, 0.0712503 , 0.98424964, 0.20258022, 0.33971214],
                [0.6561921 , 0.87046826, 0.86435258, 0.74885978, 0.60192395]])

In [38]: np.log2(ary)

Out[38]: array([[ -1.98701225, -1.95495074, -1.09006592, -0.19384527, -1.17346854],
                [-2.45718189, -1.25921923, -0.18527656, -1.44427205, -3.67954015],
                [-1.34283762, -4.77863152, -0.26140383, -1.34611384, -0.7337791 ],
                [-0.75104841, -3.8109602 , -0.02290381, -2.30343481, -1.5576153 ],
                [-0.60780988, -0.20013641, -0.21030818, -0.41723248, -0.73234687]])

```

```

In [39]: np.sin(ary)

Out[39]: array([[0.24959381, 0.25507919, 0.45265438, 0.7670769 , 0.42897173],
               [0.18109715, 0.40572324, 0.77040776, 0.35926252, 0.07796633],
               [0.38411072, 0.03642441, 0.7408102 , 0.38328489, 0.56573694],
               [0.55982254, 0.07119003, 0.83285701, 0.20119745, 0.3332157 ],
               [0.6101042 , 0.7646308 , 0.76067516, 0.68080403, 0.56622933]])

In [40]: np.sum(ary)

Out[40]: 12.415822346118937

In [43]: np.sum(ary, axis=0)

Out[43]: array([2.07897086, 1.65385061, 4.03209832, 2.5865401 , 2.06436246])

In [44]: np.max(ary)

Out[44]: 0.9842496432443214

In [45]: np.mean(ary)

Out[45]: 0.49663289384475745

In [46]: np.std(ary)

Out[46]: 0.28134296733775815

In [47]: np.var(ary)

Out[47]: 0.07915386527041485

In [49]: ary[1:,2:4]

Out[49]: array([[0.87948047, 0.36747753],
               [0.83427572, 0.39335018],
               [0.98424964, 0.20258022],
               [0.86435258, 0.74885978]])

In [51]: np.sort(ary)

Out[51]: array([[0.25226077, 0.2579296 , 0.44335414, 0.46973991, 0.87427238],
               [0.07804553, 0.18210193, 0.36747753, 0.41776999, 0.87948047],
               [0.03643247, 0.39335018, 0.39424446, 0.60132669, 0.83427572],
               [0.0712503 , 0.20258022, 0.33971214, 0.59417161, 0.98424964],
               [0.60192395, 0.6561921 , 0.74885978, 0.86435258, 0.87046826]])

In [52]: ary

```

```
Out [52]: array([[0.25226077, 0.2579296 , 0.46973991, 0.87427238, 0.44335414],
                [0.18210193, 0.41776999, 0.87948047, 0.36747753, 0.07804553],
                [0.39424446, 0.03643247, 0.83427572, 0.39335018, 0.60132669],
                [0.59417161, 0.0712503 , 0.98424964, 0.20258022, 0.33971214],
                [0.6561921 , 0.87046826, 0.86435258, 0.74885978, 0.60192395]])
```

```
In [53]: ary.T
```

```
Out [53]: array([[0.25226077, 0.18210193, 0.39424446, 0.59417161, 0.6561921 ],
                [0.2579296 , 0.41776999, 0.03643247, 0.0712503 , 0.87046826],
                [0.46973991, 0.87948047, 0.83427572, 0.98424964, 0.86435258],
                [0.87427238, 0.36747753, 0.39335018, 0.20258022, 0.74885978],
                [0.44335414, 0.07804553, 0.60132669, 0.33971214, 0.60192395]])
```

```
In [54]: ary[:3,:].T
```

```
Out [54]: array([[0.25226077, 0.18210193, 0.39424446],
                [0.2579296 , 0.41776999, 0.03643247],
                [0.46973991, 0.87948047, 0.83427572],
                [0.87427238, 0.36747753, 0.39335018],
                [0.44335414, 0.07804553, 0.60132669]])
```

```
In [55]: ary[:3,:].flatten()
```

```
Out [55]: array([0.25226077, 0.2579296 , 0.46973991, 0.87427238, 0.44335414,
                0.18210193, 0.41776999, 0.87948047, 0.36747753, 0.07804553,
                0.39424446, 0.03643247, 0.83427572, 0.39335018, 0.60132669])
```

```
In [59]: ary = np.array([4,5,6,7,8])
        ary
```

```
Out [59]: array([4, 5, 6, 7, 8])
```

```
In [61]: ary1 = np.append(ary,5)
        ary
```

```
Out [61]: array([4, 5, 6, 7, 8])
```

```
In [62]: ary2 = np.insert(ary, 0, [1,2,3])
        ary2
```

```
Out [62]: array([1, 2, 3, 4, 5, 6, 7, 8])
```

```
In [65]: ary3 = np.delete(ary2, 0)
        ary3
```

```
Out [65]: array([2, 3, 4, 5, 6, 7, 8])
```

```
In [67]: aryC = ary3.copy()
        aryC
```



```
Out[67]: array([2, 3, 4, 5, 6, 7, 8])
```

```
In [76]: ary1 = np.array([[1,2,3],[4,5,6]])  
        ary2 = np.array([[4,5,6],[7,8,9]])  
        ary3 = np.array([[8,7,6],[1,5,9]])  
        ary_cat = np.concatenate((ary1,ary2,ary3),axis=1)  
        print(ary_cat)
```

```
[[1 2 3 4 5 6 8 7 6]  
 [4 5 6 7 8 9 1 5 9]]
```

```
In [77]: catV = np.vstack((ary1, ary2,ary3))  
        catV
```

```
Out[77]: array([[1, 2, 3],  
                [4, 5, 6],  
                [4, 5, 6],  
                [7, 8, 9],  
                [8, 7, 6],  
                [1, 5, 9]])
```

```
In [78]: catH = np.hstack((ary1, ary2,ary3))  
        catH
```

```
Out[78]: array([[1, 2, 3, 4, 5, 6, 8, 7, 6],  
                [4, 5, 6, 7, 8, 9, 1, 5, 9]])
```

```
In [80]: ary = np.array([1,5,3,6,7,4,245,1,7,45,7,56,36])  
        np.unique(ary)
```

```
Out[80]: array([ 1,  3,  4,  5,  6,  7, 36, 45, 56, 245])
```

```
In [85]: #intersection, differntiation, neither  
        ary1 = np.array([1,2,3,4])  
        ary2 = np.array([2,3,6,4])  
        print(np.intersect1d(ary1, ary2))  
        print(np.union1d(ary1, ary2))  
        print(np.setdiff1d(ary2, ary1))  
        print(np.setxor1d(ary1, ary2))
```

```
[2 3 4]  
[1 2 3 4 6]  
[6]  
[1 6]
```

```
In [4]: import pandas as pd
```

```
In [88]: #for pandas
```

```
In [5]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set(color_codes=True)
```

```
In [9]: data = pd.read_csv('/home/user/GODAVARI/coOrdinate1.csv')
data.head()
```

```
Out[9]:
```

	Unnamed: 0	Albedo_inst	AvgSurfT_inst	CanopInt_inst	ECanop_tavg	
0	0	11.920000	308.211884	0.0	0.00	
1	0	11.890000	307.378479	0.0	0.00	
2	0	11.849998	307.798584	0.0	0.00	
3	0	11.820000	308.330750	0.0	0.04	
4	0	11.790000	307.115631	0.0	0.00	

	Esoil_tavg	Evap_tavg	LWdown_f_tavg	Lwnet_tavg	PotEvap_tavg	...	
0	18.570000	0.000082	340.456329	-167.500580	741.664856	...	
1	18.219999	0.000081	335.034790	-166.838684	769.109375	...	
2	16.670000	0.000078	332.093597	-170.586090	739.671814	...	
3	15.900000	0.000076	339.926392	-165.926682	739.387756	...	
4	16.160000	0.000075	331.670410	-165.583313	794.559509	...	

	SoilMoi40_100cm_inst	SoilTMP0_10cm_inst	SoilTMP100_200cm_inst	
0	154.382996	297.980682	297.915070	
1	153.203995	297.227661	297.881104	
2	152.104996	296.832367	297.839264	
3	151.041000	297.271484	297.801300	
4	149.955002	296.841766	297.758514	

	SoilTMP10_40cm_inst	SoilTMP40_100cm_inst	Swnet_tavg	Tair_f_inst	
0	295.425629	296.595581	657.469971	303.633453	
1	295.138550	296.530579	662.510010	302.845917	
2	294.825287	296.441772	665.369995	302.986023	
3	294.642761	296.326202	658.780029	303.447296	
4	294.584229	296.230713	667.059998	302.692627	

	Tveg_tavg	Wind_f_inst	geometry
0	186.119995	1.908393	<map object at 0x7f1b31fc6a90>
1	183.759995	2.500216	<map object at 0x7f1b32105780>
2	177.369995	1.803373	<map object at 0x7f1b321055f8>
3	172.919998	1.902477	<map object at 0x7f1b31ff14a8>
4	172.009995	3.501946	<map object at 0x7f1b32105a58>

[5 rows x 38 columns]

```
In [10]: data.dtypes
```

```

Out[10]: Unnamed: 0          int64
         Albedo_inst        float64
         AvgSurfT_inst       float64
         CanopInt_inst       float64
         ECanop_tavg         float64
         ESoil_tavg          float64
         Evap_tavg           float64
         LWdown_f_tavg       float64
         Lwnet_tavg          float64
         PotEvap_tavg        float64
         Psurf_f_inst        float64
         Qair_f_inst         float64
         Qg_tavg             float64
         Qh_tavg             float64
         Qle_tavg            float64
         Qs_acc              float64
         Qsb_acc             float64
         Qsm_acc             int64
         Rainf_f_tavg        float64
         Rainf_tavg          float64
         RootMoist_inst      float64
         SWE_inst            int64
         SWdown_f_tavg       float64
         SnowDepth_inst      int64
         Snowf_tavg          int64
         SoilMoi0_10cm_inst   float64
         SoilMoi100_200cm_inst float64
         SoilMoi10_40cm_inst  float64
         SoilMoi40_100cm_inst float64
         SoilTMP0_10cm_inst   float64
         SoilTMP100_200cm_inst float64
         SoilTMP10_40cm_inst  float64
         SoilTMP40_100cm_inst float64
         Swnet_tavg          float64
         Tair_f_inst         float64
         Tveg_tavg           float64
         Wind_f_inst         float64
         geometry            object
         dtype: object

```

```

In [11]: data.columns

```

```

Out[11]: Index(['Unnamed: 0', 'Albedo_inst', 'AvgSurfT_inst', 'CanopInt_inst',
               'ECanop_tavg', 'ESoil_tavg', 'Evap_tavg', 'LWdown_f_tavg', 'Lwnet_tavg',
               'PotEvap_tavg', 'Psurf_f_inst', 'Qair_f_inst', 'Qg_tavg', 'Qh_tavg',
               'Qle_tavg', 'Qs_acc', 'Qsb_acc', 'Qsm_acc', 'Rainf_f_tavg',
               'Rainf_tavg', 'RootMoist_inst', 'SWE_inst', 'SWdown_f_tavg',
               'SnowDepth_inst', 'Snowf_tavg', 'SoilMoi0_10cm_inst',

```

```

'SoilMoi100_200cm_inst', 'SoilMoi10_40cm_inst', 'SoilMoi40_100cm_inst',
'SoilTMP0_10cm_inst', 'SoilTMP100_200cm_inst', 'SoilTMP10_40cm_inst',
'SoilTMP40_100cm_inst', 'Swnet_tavg', 'Tair_f_inst', 'Tveg_tavg',
'Wind_f_inst', 'geometry'],
dtype='object')

```

```
In [13]: data.describe(include='all')
```

```

Out[13]:
      Unnamed: 0  Albedo_inst  AvgSurfT_inst  CanopInt_inst  ECanop_tavg  \
count      5843.0    5843.000000    5843.000000    5843.000000    5843.000000
unique         NaN             NaN             NaN             NaN             NaN
top            NaN             NaN             NaN             NaN             NaN
freq           NaN             NaN             NaN             NaN             NaN
mean           0.0     13.528428     312.764254      0.038741     18.791595
std            0.0     1.597749      8.716488      0.116089     48.560459
min            0.0     11.500000     294.225433      0.000000      0.000000
25%            0.0     12.320000     306.353516      0.000000      0.000000
50%            0.0     13.240000     309.525757      0.000000      0.000000
75%            0.0     14.150000     319.607880      0.001000      6.510000
max            0.0     17.000000     334.416504      0.500000     409.750000

      ESoil_tavg  Evap_tavg  LWdown_f_tavg  Lwnet_tavg  PotEvap_tavg  \
count  5843.000000  5843.000000    5843.000000    5843.000000    5843.000000
unique         NaN             NaN             NaN             NaN             NaN
top            NaN             NaN             NaN             NaN             NaN
freq           NaN             NaN             NaN             NaN             NaN
mean     66.761790      0.000077     412.222789    -127.603282     698.790264
std     63.233389      0.000062      38.467980      62.357138     231.067941
min       0.700000      0.000001     308.770874    -297.765778      25.336960
25%      12.155000      0.000019     381.936813    -176.656120     572.310974
50%      43.570000      0.000065     417.293640    -131.319382     716.754517
75%     111.680000      0.000126     441.833862     -72.739315     864.101135
max     323.299988      0.000252     518.161743      5.090859    1233.212646

      ...  SoilMoi40_100cm_inst  SoilTMP0_10cm_inst  SoilTMP100_200cm_inst  \
count  ...                5843.000000        5843.000000        5843.000000
unique  ...                  NaN                NaN                NaN
top     ...                  NaN                NaN                NaN
freq    ...                  NaN                NaN                NaN
mean    ...                166.106882        306.142527        301.507732
std     ...                56.060541         8.080839         3.100573
min     ...                104.426003        293.527222        296.317322
25%     ...                117.391998        300.149872        298.836609
50%     ...                137.955002        302.815735        301.324585
75%     ...                223.673500        312.016769        303.848770
max     ...                269.950989        325.580750        308.313293

      SoilTMP10_40cm_inst  SoilTMP40_100cm_inst  Swnet_tavg  Tair_f_inst  \

```

count	5843.000000	5843.000000	5843.000000	5843.000000
unique	NaN	NaN	NaN	NaN
top	NaN	NaN	NaN	NaN
freq	NaN	NaN	NaN	NaN
mean	301.763629	301.737592	607.622805	307.432248
std	4.918302	4.052518	157.954647	5.756286
min	292.833801	294.751099	13.850000	292.790863
25%	297.948441	298.414230	542.605011	303.316788
50%	301.063080	301.309723	639.859985	306.000793
75%	304.532776	304.418961	719.619995	311.177841
max	314.112000	311.308136	909.049988	323.115479

	Tveg_tavg	Wind_f_inst	geometry
count	5843.000000	5843.000000	5843
unique	NaN	NaN	5843
top	NaN	NaN	<map object at 0x7f1b31f9c3c8>
freq	NaN	NaN	1
mean	107.845247	2.127956	NaN
std	114.331925	0.958956	NaN
min	0.000000	0.007467	NaN
25%	10.980000	1.413756	NaN
50%	57.799999	2.101675	NaN
75%	189.979996	2.714424	NaN
max	412.649994	7.410892	NaN

[11 rows x 38 columns]

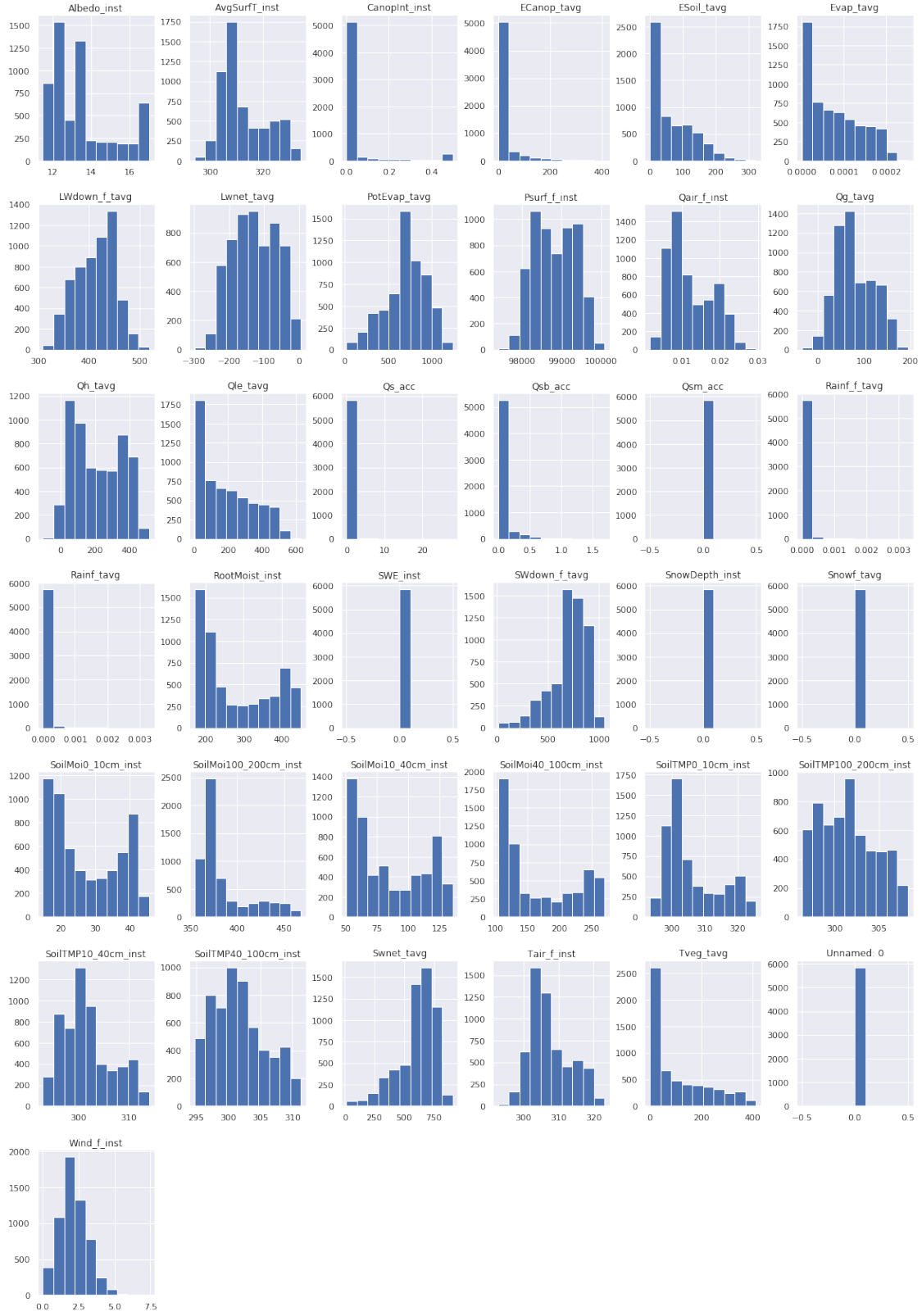
In [14]: data.hist(figsize=(20,30))

Out[14]: array([[<matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a7e290b8>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a79e3550>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a7986d68>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a79b4710>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a795a780>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a797edd8>],  
 [<matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a792d470>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a78d3c88>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a78d3cc0>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a78a4f60>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a7850908>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a787e2b0>],  
 [<matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a7821c18>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a77cf5c0>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a77f3f28>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a77a08d0>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a774f278>,  
 <matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a7774be0>],  
 [<matplotlib.axes.\_subplots.AxesSubplot object at 0x7fd6a7721588>],

```

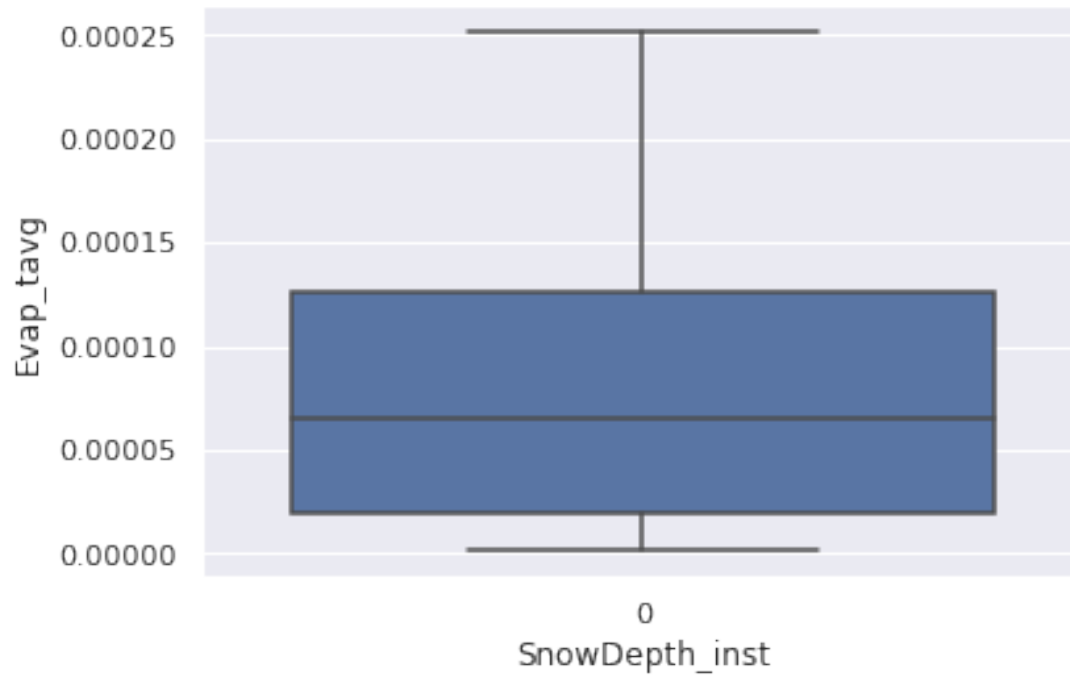
<matplotlib.axes._subplots.AxesSubplot object at 0x7fd6a76c7ef0>,
<matplotlib.axes._subplots.AxesSubplot object at 0x7fd6a76f4898>,
<matplotlib.axes._subplots.AxesSubplot object at 0x7fd6a76a1240>,
<matplotlib.axes._subplots.AxesSubplot object at 0x7fd6a7647ba8>,
<matplotlib.axes._subplots.AxesSubplot object at 0x7fd6a7675550>],
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[<matplotlib.axes._subplots.AxesSubplot object at 0x7fd6a751c828>,
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<matplotlib.axes._subplots.AxesSubplot object at 0x7fd6a7371160>]],
dtype=object)

```



```
In [15]: sns.boxplot(x="SnowDepth_inst", y="Evap_tavg", data = data)
```

```
Out[15]: <matplotlib.axes._subplots.AxesSubplot at 0x7fd6a60f0f28>
```



```
In [ ]: sns.pairplot(data)
```

```
Out[ ]: <seaborn.axisgrid.PairGrid at 0x7fd6a5fd47b8>
```

```
In [ ]:
```