

4. Visualization

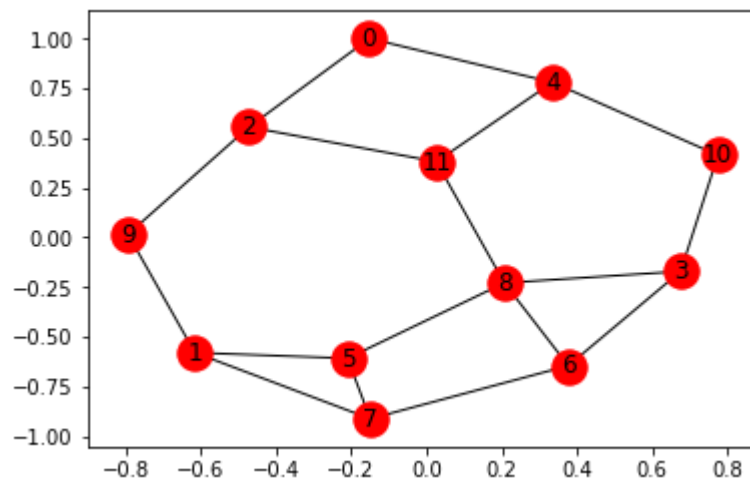
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Date : Feb. 8, 2016; Sep 9, 2019

```
In [1]: 1 %matplotlib inline
        2 import networkx as nx
        3 import matplotlib.pyplot as plt
        4 import random
```

Let's fix node positions

```
In [2]: 1 G = nx.read_gexf('netlab/data/us-att.gexf', node_type=int)
        2 layout = nx.spring_layout(G)
        3 nx.draw_networkx(G, pos=layout)
```



```
In [ ]: 1
```

Data struture of the layout

The layout data is a dictionary with

- nodes as *keys* and
- positions as *values* in array format of numpy.

We can make customized layout data. investigate "us-att" data.

In [3]: 1 G.nodes(data=True)

Out[3]: NodeDataView({0: {'latitude': 40.712756, 'abbr': 'nwy', 'name': 'New York, NY', 'longitude': -74.006047, 'population': 8175133, 'label': '0'}, 1: {'latitude': 33.94352, 'abbr': 'lax', 'name': 'Los Angeles, CA', 'longitude': -118.40866, 'population': 3792621, 'label': '1'}, 2: {'latitude': 41.878247, 'abbr': 'chi', 'name': 'Chicago, IL', 'longitude': -87.629767, 'population': 2695598, 'label': '2'}, 3: {'latitude': 29.76429, 'abbr': 'hst', 'name': 'Houston, TX', 'longitude': -95.3837, 'population': 2099451, 'label': '3'}, 4: {'latitude': 39.952622, 'abbr': 'phl', 'name': 'Philadelphia, PA', 'longitude': -75.165708, 'population': 1526006, 'label': '4'}, 5: {'latitude': 33.445412, 'abbr': 'phx', 'name': 'Phoenix, AR', 'longitude': -112.073961, 'population': 1445632, 'label': '5'}, 6: {'latitude': 29.42373, 'abbr': 'san', 'name': 'San Antonio, TX', 'longitude': -98.49438, 'population': 1327407, 'label': '6'}, 7: {'latitude': 32.715786, 'abbr': 'sdg', 'name': 'San Diego, CA', 'longitude': -117.15834, 'population': 1307402, 'label': '7'}, 8: {'latitude': 32.803468, 'abbr': 'dal', 'name': 'Dallas, TX', 'longitude': -96.769879, 'population': 1197816, 'label': '8'}, 9: {'latitude': 37.339458, 'abbr': 'sjs', 'name': 'San Jose, CA', 'longitude': -121.895022, 'population': 945942, 'label': '9'}, 10: {'latitude': 30.332428, 'abbr': 'jvk', 'name': 'Jacksonville, FL', 'longitude': -81.656165, 'population': 821784, 'label': '10'}, 11: {'latitude': 39.768663, 'abbr': 'ind', 'name': 'Indianapolis, IN', 'longitude': -86.159855, 'population': 820445, 'label': '11'}})

Atlas

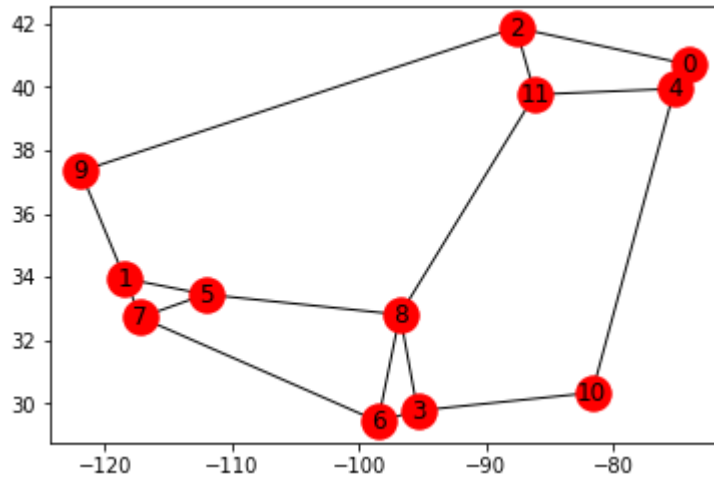
The data contains 'latitude' and 'longitude'. Let's use them for node positioning.

In [4]:

```
1 import numpy as np
2 px = nx.get_node_attributes(G, 'longitude').values()
3 py = nx.get_node_attributes(G, 'latitude').values()
4 pos = zip(px, py) # change them to a list of tuples
5 layout = dict(zip(G, pos)) # combine the node list and the position list
6 layout
```

Out[4]: {0: (-74.006047, 40.712756),
1: (-118.40866, 33.94352),
2: (-87.629767, 41.878247),
3: (-95.3837, 29.76429),
4: (-75.165708, 39.952622),
5: (-112.073961, 33.445412),
6: (-98.49438, 29.42373),
7: (-117.15834, 32.715786),
8: (-96.769879, 32.803468),
9: (-121.895022, 37.339458),
10: (-81.656165, 30.332428),
11: (-86.159855, 39.768663)}

```
In [5]: 1 nx.draw_networkx(G, pos=layout)
```



Which one is which one?

Even with the positioning, it is quite difficult to imagine the graph representing US information highway.

Why?

Because there is no information about an atlas map.

Let's draw the network on the top of US atlas map.

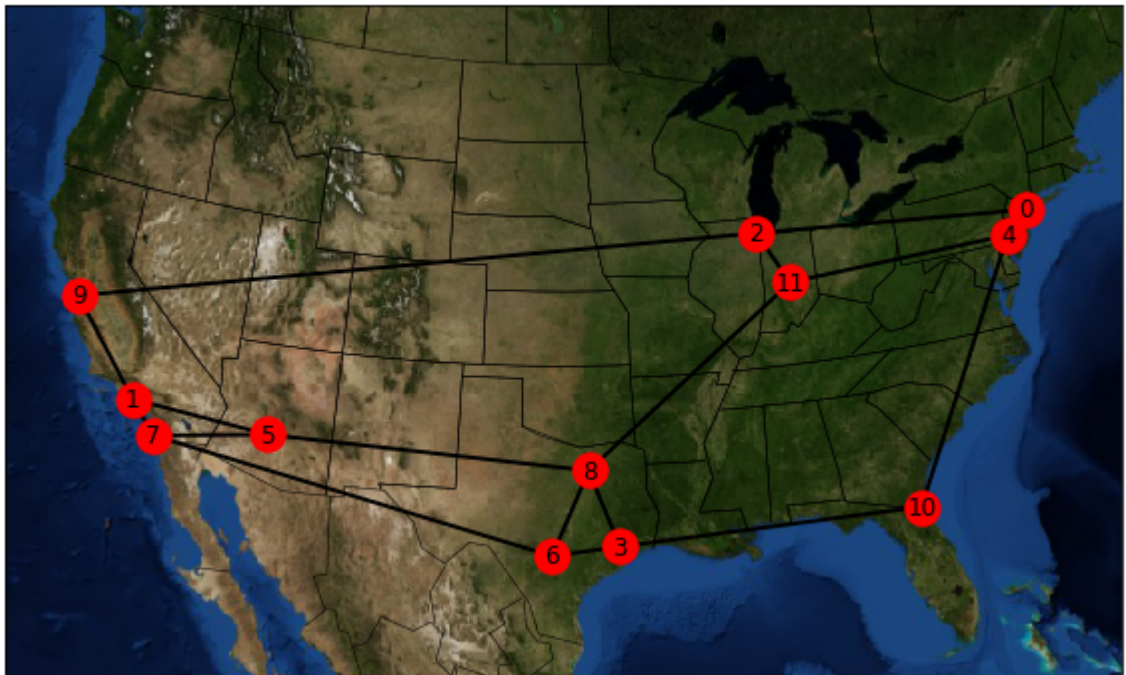
In [6]:

```

1 from mpl_toolkits.basemap import Basemap
2 import numpy as np
3 import matplotlib.pyplot as plt
4
5 plt.figure(1,figsize=(10,10))
6 lons = list(nx.get_node_attributes(G, 'longitude').values())
7 lats = list(nx.get_node_attributes(G, 'latitude').values())
8
9 m = Basemap(width=5.e6,height=3.e6,
10             projection='gnom',lat_0=38.,lon_0=-98.)
11
12 m.drawcountries()
13 m.drawstates()
14 m.bluemarble()
15
16 px, py = m(lons,lats)
17 pos = zip(px, py) # The coordination must be projected to
18 layout = dict(zip(G, pos)) # Convert the projected data to a dictionary of node I
19 nx.draw_networkx(G, pos=layout, width=2.0) # and NumPy array.
20
21 plt.title('US-ATT')
22 plt.show()

```

US-ATT



In [7]: 1 `help(plt.figure)`

Help on function figure in module matplotlib.pyplot:

```
figure(num=None, figsize=None, dpi=None, facecolor=None, edgecolor=None, frameon=True, FigureClass=<class 'matplotlib.figure.Figure'>, clear=False, **kwargs)
    Creates a new figure.
```

Parameters

`num` : integer or string, optional, default: none

If not provided, a new figure will be created, and the figure number will be incremented. The figure object holds this number in a ``number`` attribute.

If `num` is provided, and a figure with this id already exists, make it active, and returns a reference to it. If this figure does not exist, create it and returns it.

If `num` is a string, the window title will be set to this figure's ``num``.

`figsize` : tuple of integers, optional, default: None

width, height in inches. If not provided, defaults to `rc figure.figsize`.

`dpi` : integer, optional, default: None

resolution of the figure. If not provided, defaults to `rc figure.dpi`.

`facecolor` :

the background color. If not provided, defaults to `rc figure.facecolor`.

`edgecolor` :

the border color. If not provided, defaults to `rc figure.edgecolor`.

`frameon` : bool, optional, default: True

If False, suppress drawing the figure frame.

`FigureClass` : class derived from `matplotlib.figure.Figure`

Optionally use a custom Figure instance.

`clear` : bool, optional, default: False

If True and the figure already exists, then it is cleared.

Returns

`figure` : Figure

The Figure instance returned will also be passed to `new_figure_manager` in the backends, which allows to hook custom Figure classes into the pylab interface. Additional kwargs will be passed to the figure init function.

Notes

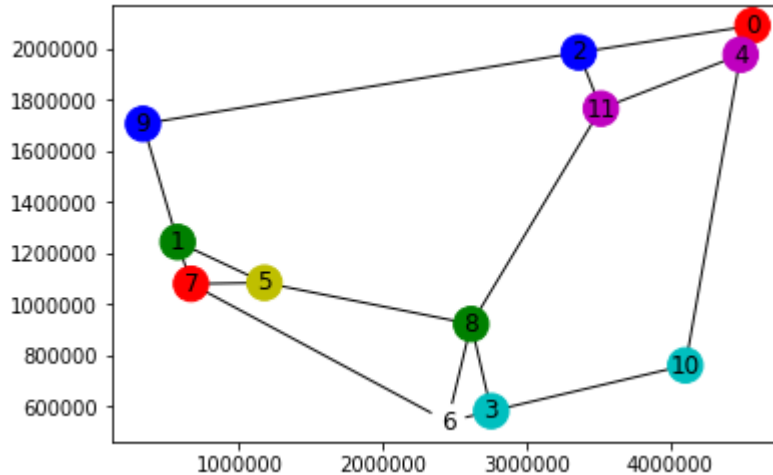
If you are creating many figures, make sure you explicitly call "close" on the figures you are not using, because this will enable pylab to properly clean up the memory.

rcParams defines the default values, which can be modified in the matplotlibrc file

Characterizing node

The node colors can be customized by the name parameter "node_color".

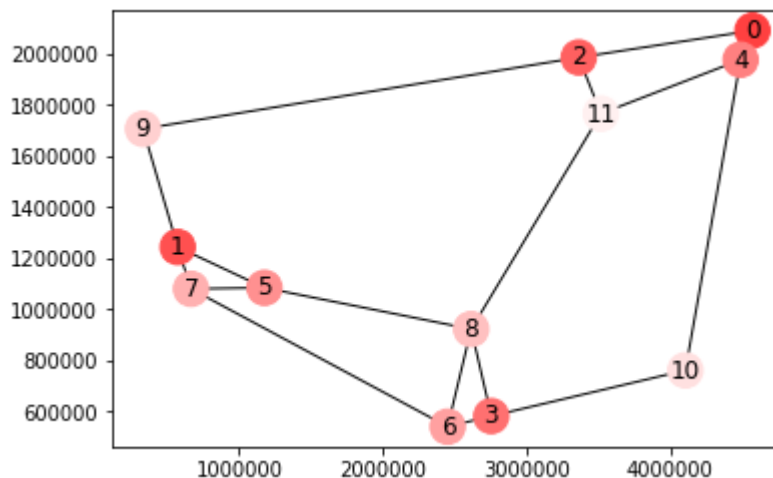
```
In [8]: 1 color = ['r', 'g', 'b', 'c', 'm', 'y', 'w', 'r', 'g', 'b', 'c', 'm']
2         nx.draw_networkx(G, pos=layout, node_color=color)
```



Fine control of the color

Moreover, the hexadecimal RGB color code can be used.

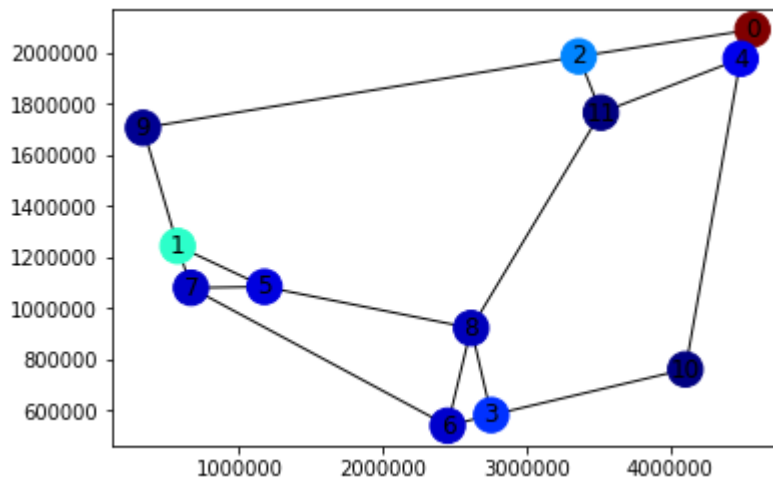
```
In [9]: 1 color = ['#FF4040', '#FF5050', '#FF6060', '#FF7070', '#FF8080', '#FF9090',
2             '#FFA0A0', '#FFB0B0', '#FFC0C0', '#FFD0D0', '#FFE0E0', '#FFF0F0']
3         nx.draw_networkx(G, pos=layout, node_color=color)
```



Visualize the "Population"

```
In [10]: 1 import matplotlib.colors as colors
2 import matplotlib.cm as cm
3
4 pop = list(nx.get_node_attributes(G, 'population').values())
5 cval = np.array(pop)
6 print (cval)
7 color_map = cm.get_cmap('jet') # [1]
8 nx.draw_networkx(G, pos=layout,
9                 node_color=cval,
10                cmap=color_map, vmin=np.min(cval), vmax=np.max(cval))
```

```
[8175133 3792621 2695598 2099451 1526006 1445632 1327407 1307402 1197816
 945942  821784  820445]
```



[1] http://matplotlib.org/examples/color/colormaps_reference.html
http://matplotlib.org/examples/color/colormaps_reference.html

Big gap between 1st and 2nd cities

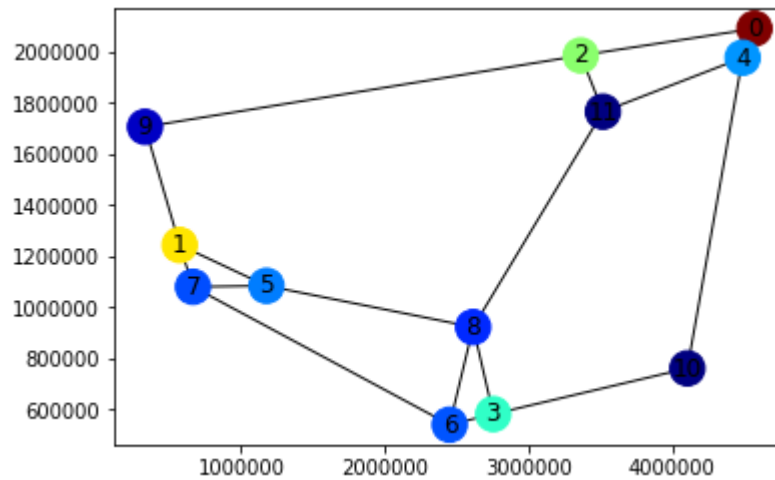
There are big jumps between

- NWY and LAX
- LAX and CHI

But, there are not much difference among the second half. Instead of the linear mapping, let's use a logarithmic mapping.

```
In [11]: 1 cval = np.log(np.array(pop))
2 print (cval)
3 nx.draw_networkx(G, pos=layout,
4                 node_color=cval,
5                 cmap=color_map, vmin=np.min(cval), vmax=np.max(cval))
```

```
[15.91660754 15.14856789 14.80713063 14.55718644 14.23816442 14.18405715
14.09873797 14.08355252 13.99601046 13.75993654 13.61923287 13.61760215]
```



Put-it-all-together

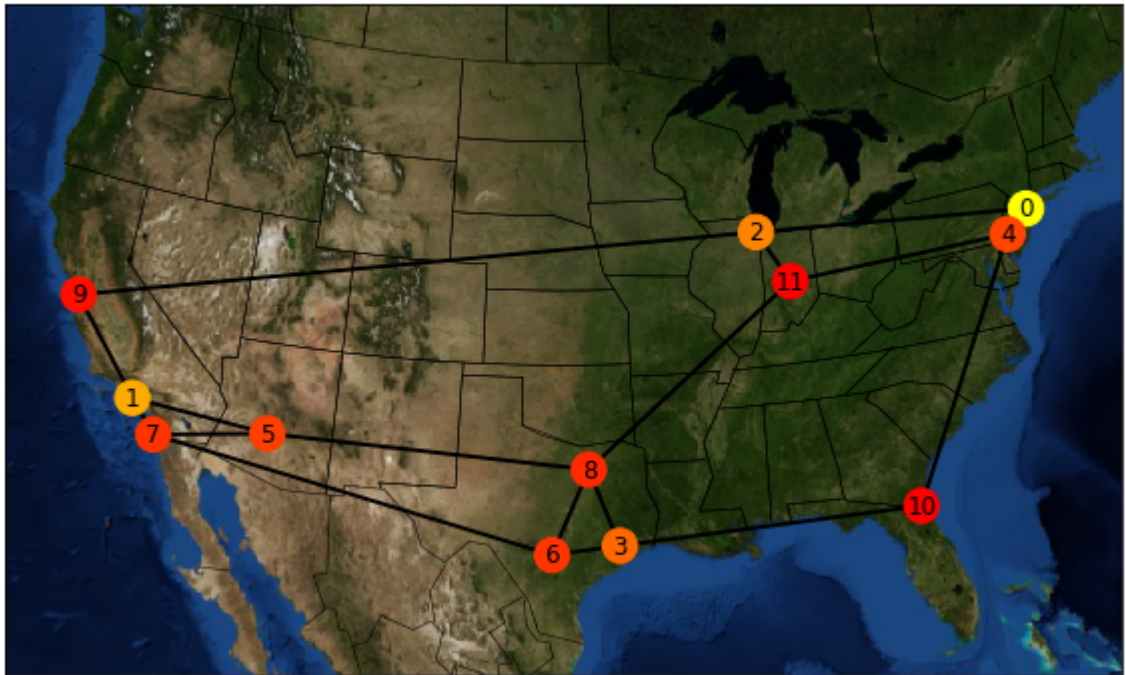
In [12]:

```

1 plt.figure(1,figsize=(10,10))
2 px = list(nx.get_node_attributes(G, 'longitude').values())
3 py = list(nx.get_node_attributes(G, 'latitude').values())
4 pop = list(nx.get_node_attributes(G, 'population').values())
5
6
7 cval = np.log(np.array(pop))
8 color_map = cm.get_cmap('autumn')
9
10 m = Basemap(width=5.e6,height=3.e6,\
11             projection='gnom',lat_0=38.,lon_0=-98.)
12
13 m.drawcountries()
14 m.drawstates()
15 m.bluemarble()
16
17 px, py = m(px,py) # The coordination must be projected to the
18 pos = zip(px, py) # Convert the projected data to a dictionary of node I
19 layout = dict(zip(G, pos)) # and NumPy array.
20
21 nx.draw_networkx(G, pos=layout,
22                 node_color=cval,
23                 cmap=color_map, vmin=np.min(cval), vmax=np.max(cval),
24                 width=2.0)
25
26 plt.title('US-ATT')
27 plt.show()

```

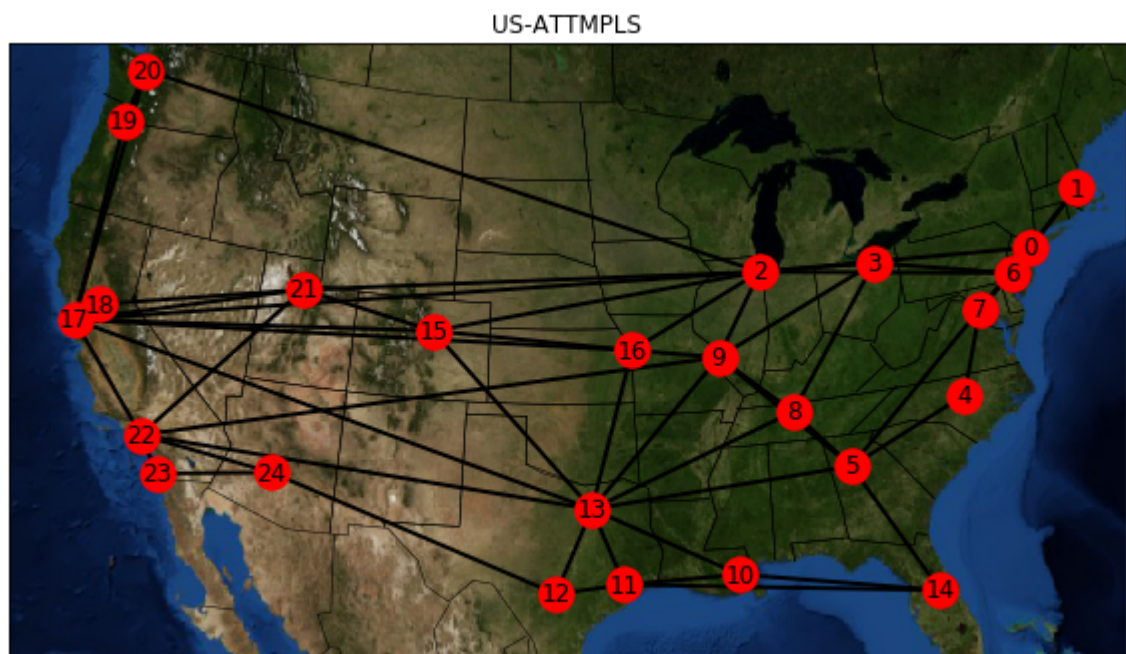
US-ATT



```

In [13]: 1 import networkx as nx
          2 import numpy as np
          3 from mpl_toolkits.basemap import Basemap
          4 import matplotlib.pyplot as plt
          5 import matplotlib.colors as colors
          6 import matplotlib.cm as cm
          7 import random
          8
          9 %matplotlib inline
         10
         11 G = nx.read_gexf('netlab/data/atmpls.gexf')
         12
         13 plt.figure(1,figsize=(10,10))
         14 lons = list(nx.get_node_attributes(G, 'longitude').values())
         15 lats = list(nx.get_node_attributes(G, 'latitude').values())
         16
         17 m = Basemap(width=5.e6,height=3.e6,\
         18             projection='gnom',lat_0=38.,lon_0=-98.)
         19
         20 m.drawcountries()
         21 m.drawstates()
         22 m.bluemarble()
         23
         24 px, py = m(lons, lats) # The coordination must be projected to
         25 pos = zip(px, py) # Convert the projected data to a dictionary of node I
         26 layout = dict(zip(G, pos))
         27
         28 nx.draw_networkx(G, pos=layout, width=2.0)
         29
         30 plt.title('US-ATMPLS')
         31 plt.show()

```



In []:

1

