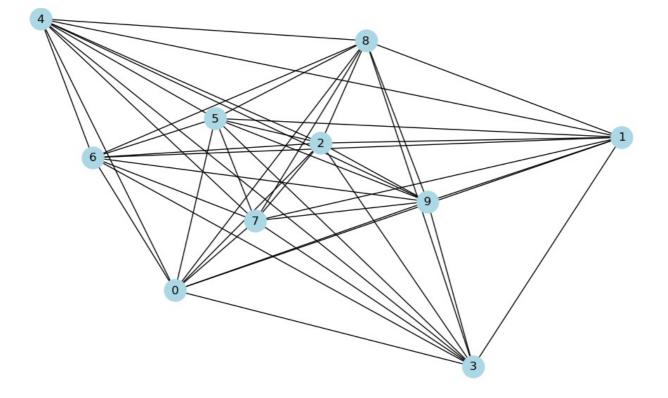
```
In [15]:
```

```
import random
import networkx as nx
import matplotlib.pyplot as plt
from itertools import combinations, groupby
import time
from tqdm import tqdm
from networkx.algorithms import tree
```

In [16]:

```
def gnp random connected graph (num of nodes: int,
                                completeness: int,
                                directed: bool = False,
                                draw: bool = False):
    Generates a random graph, similarly to an Erdős-Rényi
    graph, but enforcing that the resulting graph is conneted (in case of undirected grap
hs)
    .....
    if directed:
       G = nx.DiGraph()
    else:
       G = nx.Graph()
    edges = combinations(range(num of nodes), 2)
    G.add nodes from(range(num of nodes))
    for , node edges in groupby(edges, key = lambda x: x[0]):
        node edges = list(node edges)
        random edge = random.choice(node edges)
        if random.random() < 0.5:</pre>
            random edge = random edge[::-1]
        G.add edge(*random edge)
        for e in node edges:
            if random.random() < completeness:</pre>
                G.add edge(*e)
    for (u,v,w) in G.edges(data=True):
        w['weight'] = random.randint(-5, 20)
        \# w = random.randint(-5, 20)
    if draw:
        plt.figure(figsize=(10,6))
        if directed:
            # draw with edge weights
            pos = nx.arf layout(G)
            nx.draw(G,pos, node color='lightblue',
                    with labels=True,
                    node size=500,
                    arrowsize=20,
                    arrows=True)
            labels = nx.get_edge_attributes(G,'weight')
            nx.draw networkx edge labels(G, pos,edge labels=labels)
        else:
            nx.draw(G, node color='lightblue',
                with labels=True,
                node size=500)
    return G
G = gnp random connected graph(10, 1, False, True)
```



Павлосюк Роман Коваль Вікторія Мета: реалізувати алгоритми, вивчені на лекціях, та порівняти їх з вдудованими Нижче написані алгоритми Прима та Краскала

In [17]:

```
def kruskals algorithm(graph: nx.classes.graph.Graph) -> list:
    '''Kruskal's algorithm implementation'''
   edges = sorted([(a,b,c['weight']) for a,b,c in list(graph.edges(data=True))], key= 1
ambda x: x[2])
   mst kraskal = []
   nodes set = disjoint set(list(graph.nodes))
              in edges:
    for v,u,
       if find(nodes set, v) != find(nodes set, u):
            mst kraskal.append((v, u))
            union(nodes set, v, u)
    return mst kraskal
def find(nodes set: dict, vertex: int):
    if nodes set[vertex] != vertex:
       nodes set[vertex] = find(nodes set, nodes set[vertex])
    return nodes set[vertex]
def disjoint set(list of vertices: list):
    return {k:k for k in list_of_vertices}
def union(nodes set: dict, vertex1:int, vertex2:int):
   nodes set[find(nodes set, vertex1)] = find(nodes set, vertex2)
```

In [18]:

```
visited.add(v)
    edges.remove((u,v,_))
    break

if (v in visited and u not in visited):
    prim_mst.append((u, v))
    visited.add(u)
    edges.remove((u,v,_))
    break

return prim_mst
```

In [29]:

```
list_num_nodes = [10, 20, 50, 100, 200]
# 1 completeness:
my_krusk_1 = [0.000095, 0.000347, 0.002031, 0.013096, 0.061401]
my_prim_1 = [0.000056, 0.000284, 0.001650, 0.011220, 0.055621]
notmy_krusk_1 = [0.000181, 0.000607, 0.003195, 0.014495, 0.075110]
notmy_prim_1 = [0.000084, 0.000286, 0.001445, 0.00738, 0.043895]
# 0.4 completeness
my_krusk_0 = [0.000053, 0.000196, 0.000865, 0.004372, 0.024299]
my_prim_0 = [0.000053, 0.000163, 0.000683, 0.003535, 0.021788]
notmy_krusk_0 = [0.000123, 0.000483, 0.001881, 0.005845, 0.025764]
notmy_prim_0 = [0.000072, 0.000225, 0.000872, 0.002903, 0.014101]
```

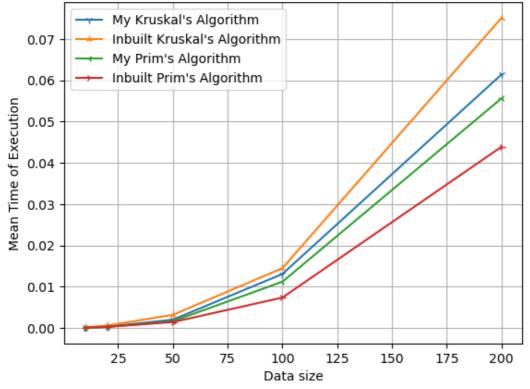
In [35]:

```
plt.plot(list_num_nodes, my_krusk_1, marker='1', label="My Kruskal's Algorithm")
plt.plot(list_num_nodes, notmy_krusk_1, marker='2', label="Inbuilt Kruskal's Algorithm")
plt.plot(list_num_nodes, my_prim_1, marker='3', label="My Prim's Algorithm")
plt.plot(list_num_nodes, notmy_prim_1, marker='4', label="Inbuilt Prim's Algorithm")

plt.xlabel('Data size')
plt.ylabel('Mean Time of Execution')
plt.title('Comparison of Algorithms on 1 Completeness')
plt.legend()

plt.grid(True)
plt.show()
```





Як видно на графіку, мій алгоритм Краскала є кращим за вбудований. Я використовував **find-union method** для перевірки на цикл. Загальна складність мого алгоритму **V(n)**. Але водночас вбудований алгоритм прима в рази швидший, особливо це помітно зі збільшенням к-сть вершин та заповненості графа.

Висновок: найкраще і найшвидше працює вбудований алгоритм Прима і загалом він є кращим за Краскала.

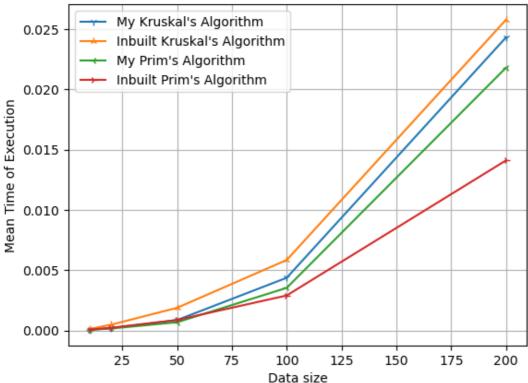
In [37]:

```
plt.plot(list_num_nodes, my_krusk_0, marker='1', label="My Kruskal's Algorithm")
plt.plot(list_num_nodes, notmy_krusk_0, marker='2', label="Inbuilt Kruskal's Algorithm")
plt.plot(list_num_nodes, my_prim_0, marker='3', label="My Prim's Algorithm")
plt.plot(list_num_nodes, notmy_prim_0, marker='4', label="Inbuilt Prim's Algorithm")

plt.xlabel('Data size')
plt.ylabel('Mean Time of Execution')
plt.title('Comparison of Algorithms on 0.4 Completeness')
plt.legend()

plt.grid(True)
plt.show()
```

Comparison of Algorithms on 0.4 Completeness



In []:

```
NUM OF ITERATIONS = 1000
 time taken = 0
#
 lst = []
 for num node in list num nodes:
      for i in tqdm(range(NUM OF ITERATIONS)):
          # note that we should not measure time of graph creation
         G = gnp random connected graph (num node, 0.4, False)
         start = time.time()
          # tree.minimum spanning tree(G, algorithm="kruskal")
         tree.minimum spanning tree(G, algorithm="prim")
          # kruskals algorithm(G)
          # prims algorithm(G)
         end = time.time()
          time taken += end - start
     lst.append(round(time taken / NUM OF ITERATIONS, 6))
 print(lst)
```