Last name	
First name	

## LARSON—OPER 731—HOMEWORK WORKSHEET 10 König-Egervary Theorem in Sage.

- 1. Log in to your Sage Cloud account.
  - (a) Start Chrome browser.
  - (b) Go to http://cocalc.com
  - (c) Click "Sign In".
  - (d) Click the project for our course.
  - (e) Click "New", call it **h10**, then click "Sage Worksheet".
- 2. The König-Egervary Theorem is a theorem for bipartite graphs. First we'll make a random bipartite graph g with 9 vertices, relabel it, and view/show it.

```
g=graphs.RandomBipartite(4,5,0.5)
g.relabel()
g.show()
```

3. Define a list V of vertices and E of edges for your graph g:

```
V = g.vertices()
E = g.edges(labels = False)
```

- 4. Evaluate V and E to see what you have.
- 5. Find a maximum matching of g (by hand) for your graph g (it will be different for each student).
- 6. Find a minimum vertex cover of graph g.

IP.get\_values(x)

- 7. Prove that your matching is maximum and your vertex cover is minimum.
- 8. Enter and run this Integer Programming model of the maximum matching problem:

```
IP = MixedIntegerLinearProgram(maximization=True)
x = IP.new_variable(integer=True,nonnegative=True)

IP.set_objective(sum(x[e] for e in E))
#sets a decision variable corresponding to each edge

for v in V: #vertex constraints (one for each vertex with an incident edge)
    IncidentEdges = [e for e in E if e[0]==v or e[1]==v]
    if len(IncidentEdges) > 0:
        IP.add_constraint(sum(x[e] for e in IncidentEdges) <= 1)

IP.solve()</pre>
```

10. Enter and run this Linear Programming relaxation of the maximum matching IP:

```
LP = MixedIntegerLinearProgram(maximization=True)
x = LP.new_variable(nonnegative=True)
LP.set_objective(sum(x[e] for e in E))

for v in V:
    IncidentEdges = [e for e in E if e[0]==v or e[1]==v]
    if len(IncidentEdges) > 0:
        LP.add_constraint(sum(x[e] for e in IncidentEdges) <= 1)

LP.solve()
LP.get_values(x)</pre>
```

- 11. We're solving an LP with real-valued variables. But what do you notice about the decision variables in the solution dictionary? Why shouldn't this surprise you (based on what we learned about this situation in class)?
- 12. Now let's set-up and solve the Dual IP. We showed in class that this models the minimum vertex cover problem. Enter and run:

```
DualIP = MixedIntegerLinearProgram(maximization=False)
y = DualIP.new_variable(integer=True, nonnegative=True)
DualIP.set_objective(sum(y[v] for v in V))
#sets a decision variable corresponding to each vertex

for e in E: #one constraint for each edge
    DualIP.add_constraint(y[e[0]]+y[e[1]] >= 1)

DualIP.solve()
DualIP.get_values(y)
```

- 13. Explain Sage's output. How many vertices are in a minimum vertex cover? What minimum vertex cover does it produce?
- 14. Now let's relax this:

```
DualLP = MixedIntegerLinearProgram(maximization=False)
y = DualLP.new_variable(nonnegative=True)
DualLP.set_objective(sum(y[v] for v in V))

for e in E:
    DualLP.add_constraint(y[e[0]]+y[e[1]] >= 1)

DualLP.solve()
DualLP.get_values(y)
```

15. Again we're solving an LP with real-valued variables. But what do you notice about the decision variables in the solution dictionary? Why shouldn't this surprise you?