```
#Troy Krupinski
#tsk0064
import networkx as nx
{\tt import\ matplotlib.pyplot\ as\ plt}
import numpy as np
G = nx.DiGraph()
#CSCE 5215 - Machine Learning
#Project 3 - Bayesian Networks
#Network:
-----
***********
Layer 4:
-----
car_won't_start:
Parents:[No_oil, Battery_flat]
children:[]
Co-parents:[]
Markov-blanket: [No_oil, Battery_flat]
Dipstick:
Parents: [No_oil]
children: []
Co-parents:[]
Markov-blanket: [No_oil]
_____
Gas_gauge:
Parents: [Battery_flat]
Children: []
Co-parents:[]
Markov-blanket: [Battery_flat]
-----
Lights:
Parents: [Battery_flat]
Children: []
Co-parents:[]
Markov-blanket: [Battery_flat]
*******
Layer 3:
-----
Battery_flat:
```

Parents: [No_charging, Battery_dead]

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Children: [car_won't_start, Gas_gauge, Lights]
Co-parents: [No_oil]
Markov-blanket: [No_charging, Battery_dead, car_won't_start, Gas_gauge, Lights, No_oil]
No_oil:
parents: []
children: [Dipstick, car_won't_start]
Co-parents: [Battery_flat]
Markov-blanket: [Dipstick, car_won't_start, Battery_flat]
*******
Layer 2:
No_charging:
Parents: [Alternator_broken, Fanbelt_broken]
Children: [Battery_flat]
Co-parents: [Battery_dead]
Markov-blanket: [Alternator_broken, Fanbelt_broken, Battery_flat, Battery_dead]
Battery_dead:
Parents: [Battery_age]
Children: [Battery_flat]
Co-parents: [No_charging]
Markov-blanket: [Battery_age, Battery_flat, No_charging]
*******
Layer 1:
Battery_age:
Parents: []
Children: [Battery_dead]
Co-parents: []
Markov-blanket: [Battery_dead]
Alternator_broken:
Parents: []
Children: [No_charging]
Co-parents: [Fanbelt_broken]
Markov-blanket: [No_charging, Fanbelt_broken]
```

```
Fanbelt_broken:
Parents: []
Children: [No_charging]
Co-parents: [Alternator_broken)
Markov-blanket: [No_charging, Alternator_broken]
G.add_node("battery_age", ba_y=.2, ba_n = .8)
G.add_node("alternator_broken", ab_y=.1, ab_n = .9)
G.add_node("fanbelt_broken", fb_y = .3, fb_n = .7)
\label{eq:gamma_decomposition} G.add\_node("battery\_dead", ba\_y\_bd\_y = .7, ba\_y\_bd\_n = .3, ba\_n\_bd\_y = .3, ba\_n\_bd\_n = .7)
G.add_node("no_charging_table",
           ab_y_fb_y_nc_y = .75,
           ab_y_fb_n_c_y = .4,
           ab_n_fb_y_nc_y = .6,
           ab_n_fb_n_c_y = .1,
           ab_y_fb_y_nc_n= 0.25,
           ab_y_fb_n_c = .6,
           ab_n_fb_y_nc_n = .4,
           ab_n_fb_n_nc_n = .9
G.add_node("battery_flat",
    bd_y_nc_y_bf_y=0.95, bd_y_nc_n_bf_y=0.85,
    bd\_n\_nc\_y\_bf\_y=0.8, \ bd\_n\_nc\_n\_bf\_y=0.1, \\
    bd_y_nc_y_bf_n=0.05, bd_y_nc_n_bf_n=0.15,
    bd_n_nc_y_bf_n=0.2, bd_n_nc_n_bf_n=0.9
)
G.add_node("no_oil", no_y=0.05, no_n=0.95)
G.add_node("lights",
    l_y_bf_y=0.9, l_n_bf_y=0.1,
    l_y_bf_n=0.3, l_n_bf_n=0.7
G.add_node("gas_gauge",
    bf_y_gg_y=0.1, bf_n_gg_y=0.95,
    bf_y_gg_n=0.9, bf_n_gg_n=0.05
G.add_node("car_wont_start",
    bf_y_no_y_cs_n=0.9, bf_y_no_n_cs_n=0.9,
    bf_n_no_y_cs_n=0.9, bf_n_no_n_cs_n=0.1,
    bf_y_no_y_cs_y=0.1, bf_y_no_n_cs_y=0.1,
    bf\_n\_no\_y\_cs\_y=0.1, \ bf\_n\_no\_n\_cs\_y=0.9
)
G.add_node("dipstick_low",
    no_y_dl_y=0.95, no_n_dl_y=0.3,
    no_y_dl_n=0.05, no_n_dl_n=0.7
edges = [
    ("battery_age", "battery_dead"),
    ("alternator_broken", "no_charging_table"),
    ("fanbelt_broken", "no_charging_table"),
    ("no_charging_table", "battery_flat"),
   ("battery_dead", "battery_flat"),
("battery_flat", "lights"),
("battery_flat", "gas_gauge"),
("battery_flat", "car_wont_start"),
    ("no_oil", "dipstick_low"),
    ("no_oil", "car_wont_start")
G.add_edges_from(edges)
plt.figure(figsize=(12, 8))
pos = nx.spring_layout(G, k=1, iterations=50)
```

```
plt.title("Bayesian Network - Project 3")
plt.show()
fanbelt_probs = G.nodes["fanbelt_broken"]
battery age probs = G.nodes["battery age"]
alternator_probs = G.nodes["alternator_broken"]
battery_dead_probs = G.nodes["battery_dead"]
no_charging_probs = G.nodes["no_charging_table"]
battery_flat_probs = G.nodes["battery_flat"]
no_oil_probs = G.nodes["no_oil"]
lights_probs = G.nodes["lights"]
gas_gauge_probs = G.nodes["gas_gauge"]
car_wont_start_probs = G.nodes["car_wont_start"]
dipstick_low_probs = G.nodes["dipstick_low"]
P(B|+j, +m) =
P(B, e, a, +j, +m)
        P(+j, +m)
        Let us take P(B, e, a, +j, +m).
               Now P(B, e, a, +j, +m) = \sum e_i a_i P(B, e, a, +j, +m) = \sum P(B) \times P(e) \times P(a|B, e) \times P(+j|a) \times P(+m|a) e, a
                P(+m|-a) \ + \ P(B) \ \times \ P(-e) \ \times \ P(+a|B, \ -e) \ \times \ P(+j|+a) \ \times \ P(+m|+a) \ + \ P(B) \ \times \ P(-e) \ \times \ P(-a|B, \ -e) \ \times \ P(-a|B, \ -e
               P(+j|-a) \times P(+m|-a)
def car_fanbelt(G):
        Calculate P(+cws|+fb) following the sum-product algorithm structure.
        P(+cws|+fb) = P(+cws,+fb) / P(+fb)
        where P(+cws,+fb) = \sum (bd,nc) P(+fb) \times P(bd) \times P(nc|+fb) \times P(+cws|bd,nc)
        # Root node probabilities
        fb_y = G.nodes["fanbelt_broken"]["fb_y"] # P(+fb)
        # Get probabilities for battery dead from battery age
        ba_y = G.nodes["battery_age"]["ba_y"]
        ba_n = G.nodes["battery_age"]["ba_n"]
        ba_y_bd_y = G.nodes["battery_dead"]["ba_y_bd_y"]
        ba_n_bd_y = G.nodes["battery_dead"]["ba_n_bd_y"]
        # Get no charging probabilities given fanbelt
        ab_y = G.nodes["alternator_broken"]["ab_y"]
        ab_n = G.nodes["alternator_broken"]["ab_n"]
        ab_y_fb_y_nc_y = G.nodes["no_charging_table"]["ab_y_fb_y_nc_y"]
        ab_n_fb_y_nc_y = G.nodes["no_charging_table"]["ab_n_fb_y_nc_y"]
        # Get no oil probabilities
        no_y = G.nodes["no_oil"]["no_y"]
        no_n = G.nodes["no_oil"]["no_n"]
        # Get battery flat probabilities
        bd_y_nc_y_bf_y = G.nodes["battery_flat"]["bd_y_nc_y_bf_y"]
        bd_y_nc_n_bf_y = G.nodes["battery_flat"]["bd_y_nc_n_bf_y"]
        bd_n_nc_y_bf_y = G.nodes["battery_flat"]["bd_n_nc_y_bf_y"]
        bd_n_nc_n_bf_y = G.nodes["battery_flat"]["bd_n_nc_n_bf_y"]
        # Get car won't start probabilities
        bf_y_no_y_cs_y = G.nodes["car_wont_start"]["bf_y_no_y_cs_y"]
        bf_y_no_n_cs_y = G.nodes["car_wont_start"]["bf_y_no_n_cs_y"]
        bf_n_no_y_cs_y = G.nodes["car_wont_start"]["bf_n_no_y_cs_y"]
        bf_n_no_n_cs_y = G.nodes["car_wont_start"]["bf_n_no_n_cs_y"]
        # Calculate P(+cws,+fb) by summing over all combinations of intermediate variables
        \# First, calculate P(bd) for both states
        p_bd_y = ba_y_bd_y * ba_y + ba_n_bd_y * ba_n # P(+bd)
        p_bd_n = 1 - p_bd_y # P(-bd)
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# Calculate P(nc|+fb) for both states
   p_nc_ngiven_fb = 1 - p_nc_ygiven_fb # P(-nc|+fb)
   # Now sum over all combinations of bd and nc states:
   # P(+cws,+fb) = \sum (bd,nc) P(+fb) \times P(bd) \times P(nc|+fb) \times P(bf|bd,nc) \times P(+cws|bf,no)
   p_cws_fb = 0
   # For battery dead = yes
   for bd_state in [(True, p_bd_y), (False, p_bd_n)]:
       bd_val, p_bd = bd_state
       # For no charging = yes/no
       for nc_state in [(True, p_nc_y_given_fb), (False, p_nc_n_given_fb)]:
           nc_val, p_nc = nc_state
           # Calculate P(bf|bd,nc)
           if bd_val and nc_val:
               p_bf_y = bd_y_nc_y_bf_y
           elif bd_val and not nc_val:
               p_bf_y = bd_y_nc_n_bf_y
           elif not bd val and nc val:
               p_bf_y = bd_n_nc_y_bf_y
           else:
               p_bf_y = bd_n_nc_n_bf_y
           p_bf_n = 1 - p_bf_y
           # Sum over battery flat states
           for bf_state in [(True, p_bf_y), (False, p_bf_n)]:
               bf_val, p_bf = bf_state
               # Sum over no oil states
               for no_state in [(True, no_y), (False, no_n)]:
                   no_val, p_no = no_state
                   # Get P(+cws|bf,no)
                   if bf val and no val:
                       p_cws = bf_y_no_y_cs_y
                   elif bf_val and not no_val:
                      p_cws = bf_y_no_n_cs_y
                   elif not bf_val and no_val:
                      p_cws = bf_n_no_y_cs_y
                   else:
                       p_cws = bf_n_no_n_cs_y
                   # Add this combination's contribution
                   p_cws_fb += fb_y * p_bd * p_nc * p_bf * p_no * p_cws
   # Finally, P(+cws|+fb) = P(+cws,+fb) / P(+fb)
   p_cws_given_fb = p_cws_fb / fb_y
   return p cws given fb
def car_battery_age(G):
   Calculate P(+cws|+ba) following the sum-product algorithm structure.
   For our case:
   P(+cws|+ba) = P(+cws,+ba) / P(+ba)
   # Root node probabilities
   ba_y = G.nodes["battery_age"]["ba_y"] # P(+ba)
   # Get probabilities for battery dead given battery age
   ba_y_bd_y = G.nodes["battery_dead"]["ba_y_bd_y"] # P(+bd|+ba)
   # Get alternator and fanbelt probabilities (root nodes)
   ab_y = G.nodes["alternator_broken"]["ab_y"]
   ab_n = G.nodes["alternator_broken"]["ab_n"]
   fb_y = G.nodes["fanbelt_broken"]["fb_y"]
   fb_n = G.nodes["fanbelt_broken"]["fb_n"]
   # Get no charging probabilities for all combinations
   ab_y_fb_y_nc_y = G.nodes["no_charging_table"]["ab_y_fb_y_nc_y"]
   ab_y_fb_n_nc_y = G.nodes["no_charging_table"]["ab_y_fb_n_nc_y"]
   ah n fh v nc v = G.nodes["no charging tahle"]["ah n fh v nc v"]
```

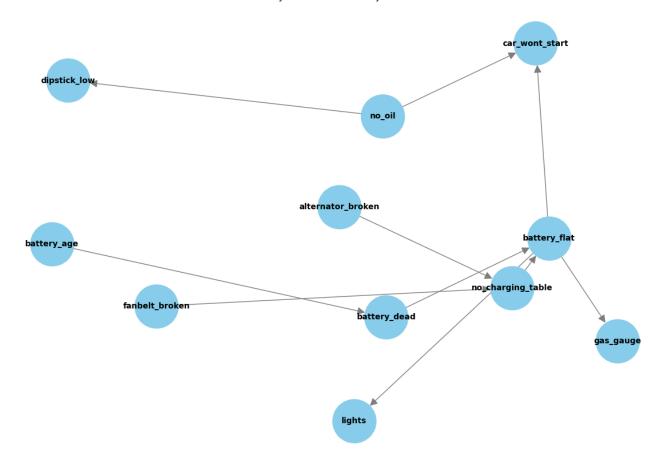
```
ab_n_fb_n_nc_y = G.nodes["no_charging_table"]["ab_n_fb_n_nc_y"]
# Get no oil probabilities (root node)
no_y = G.nodes["no_oil"]["no_y"]
no_n = G.nodes["no_oil"]["no_n"]
# Get battery flat probabilities
bd_y_nc_y_bf_y = G.nodes["battery_flat"]["bd_y_nc_y_bf_y"]
bd_y_nc_n_bf_y = G.nodes["battery_flat"]["bd_y_nc_n_bf_y"]
bd_n_nc_y_bf_y = G.nodes["battery_flat"]["bd_n_nc_y_bf_y"]
bd_n_nc_n_bf_y = G.nodes["battery_flat"]["bd_n_nc_n_bf_y"]
# Get car won't start probabilities
bf_y_no_y_cs_y = G.nodes["car_wont_start"]["bf_y_no_y_cs_y"]
bf_y_no_n_cs_y = G.nodes["car_wont_start"]["bf_y_no_n_cs_y"]
bf_n_no_y_cs_y = G.nodes["car_wont_start"]["bf_n_no_y_cs_y"]
bf_n_no_n_cs_y = G.nodes["car_wont_start"]["bf_n_no_n_cs_y"]
# Initialize probability sum
p cws ba = 0
# Given +ba, we know P(+bd|+ba)
p_bd_y = ba_y_bd_y # P(+bd|+ba)
p_bd_n = 1 - p_bd_y # P(-bd|+ba)
# Sum over all possible combinations of alternator and fanbelt states
for ab_state in [(True, ab_y), (False, ab_n)]:
    ab_val, p_ab = ab_state
    for fb_state in [(True, fb_y), (False, fb_n)]:
       fb_val, p_fb = fb_state
       # Calculate P(nc|ab,fb)
       if ab_val and fb_val:
           p_nc_y = ab_y_fb_y_nc_y
       elif ab_val and not fb_val:
           p_nc_y = ab_y_fb_n_nc_y
       elif not ab_val and fb_val:
          p_nc_y = ab_n_fb_y_nc_y
       else:
           p_nc_y = ab_n_fb_n_nc_y
       p_nc_n = 1 - p_nc_y
       # Sum over no charging states
       for nc_state in [(True, p_nc_y), (False, p_nc_n)]:
           nc_val, p_nc = nc_state
           \# Calculate P(bf|bd,nc) for both battery dead states
           for bd_state in [(True, p_bd_y), (False, p_bd_n)]:
               bd_val, p_bd = bd_state
               # Get P(bf|bd,nc)
               if bd_val and nc_val:
                   p_bf_y = bd_y_nc_y_bf_y
               elif bd_val and not nc_val:
                   p_bf_y = bd_y_nc_n_bf_y
               elif not bd_val and nc_val:
                   p_bf_y = bd_n_nc_y_bf_y
               else:
                   p_bf_y = bd_n_nc_n_bf_y
               p_bf_n = 1 - p_bf_y
               # Sum over battery flat states
               for bf_state in [(True, p_bf_y), (False, p_bf_n)]:
                   bf_val, p_bf = bf_state
                   # Sum over no oil states
                   for no_state in [(True, no_y), (False, no_n)]:
                       no_val, p_no = no_state
                       # Get P(+cws|bf,no)
                       if bf_val and no_val:
                           p_cws = bf_y_no_y_cs_y
                       elif bf_val and not no_val:
                           p_cws = bf_y_no_n_cs_y
                       elif not bf_val and no_val:
                           p_cws = bf_n_no_y_cs_y
```

```
else:
                                p_cws = bf_n_no_n_cs_y
                            # Add this combination's contribution
                             \# \ P(+cws,+ba) = P(+ba) \times P(bd|+ba) \times P(ab) \times P(fb) \times P(fb) \times P(fb|bd,nc) \times P(+cws|bf,no) \times P(no) 
                            p_cws_ba += ba_y * p_bd * p_ab * p_fb * p_nc * p_bf * p_cws * p_no
   # Finally, P(+cws|+ba) = P(+cws,+ba) / P(+ba)
   p_cws_given_ba = p_cws_ba / ba_y
   return p_cws_given_ba
def calc_prob_alternator_given_lights_gasgauge(G):
   Calculate P(+ab|-1,-gg) using Bayes' rule:
   P(+ab|-1,-gg) = P(-1,-gg|+ab)P(+ab) / P(-1,-gg)
   # Get root probabilities
    ab_y = G.nodes["alternator_broken"]["ab_y"] # P(+ab)
   ab_n = G.nodes["alternator_broken"]["ab_n"] # P(-ab)
   fb_y = G.nodes["fanbelt_broken"]["fb_y"] # P(+fb)
   fb_n = G.nodes["fanbelt_broken"]["fb_n"] # P(-fb)
   ba_y = G.nodes["battery_age"]["ba_y"] # P(+ba)
   ba_n = G.nodes["battery_age"]["ba_n"] # P(-ba)
   # Get conditional probabilities
   # Battery dead given battery age
   ba_y_bd_y = G.nodes["battery_dead"]["ba_y_bd_y"]
   ba_n_bd_y = G.nodes["battery_dead"]["ba_n_bd_y"]
   # No charging given alternator and fanbelt
   ab_y_fb_y_nc_y = G.nodes["no_charging_table"]["ab_y_fb_y_nc_y"]
   ab_y_fb_n_nc_y = G.nodes["no_charging_table"]["ab_y_fb_n_nc_y"]
    ab_n_fb_y_nc_y = G.nodes["no_charging_table"]["ab_n_fb_y_nc_y"]
   ab_n_fb_n_nc_y = G.nodes["no_charging_table"]["ab_n_fb_n_nc_y"]
   # Battery flat given battery dead and no charging
   bd_y_nc_y_bf_y = G.nodes["battery_flat"]["bd_y_nc_y_bf_y"]
   bd_y_nc_n_bf_y = G.nodes["battery_flat"]["bd_y_nc_n_bf_y"]
   bd_n_nc_y_bf_y = G.nodes["battery_flat"]["bd_n_nc_y_bf_y"]
   bd_n_nc_n_bf_y = G.nodes["battery_flat"]["bd_n_nc_n_bf_y"]
   # Lights and gas gauge given battery flat
   l_n_bf_y = G.nodes["lights"]["l_n_bf_y"] # P(-1|+bf)
   l_n_bf_n = G.nodes["lights"]["l_n_bf_n"] # P(-1|-bf)
   bf_y_gg_n = G.nodes["gas_gauge"]["bf_y_gg_n"] # P(-gg|+bf)
   bf_n_gg_n = G.nodes["gas_gauge"]["bf_n_gg_n"] # P(-gg|-bf)
   # Calculate P(-1,-gg|+ab) and P(-1,-gg|-ab)
   p_evidence_given_ab_y = 0
   p_evidence_given_ab_n = 0
    # Sum over all possible paths
    for fb_state in [(True, fb_y), (False, fb_n)]:
        fb_val, p_fb = fb_state
        # Calculate P(nc|ab,fb)
        if fb val:
            p_nc_y_given_ab_y = ab_y_fb_y_nc_y
            p_nc_y_given_ab_n = ab_n_fb_y_nc_y
            p_nc_y_given_ab_y = ab_y_fb_n_nc_y
            p_nc_y=iven_ab_n = ab_n_fb_n_nc_y
        for ba_state in [(True, ba_y), (False, ba_n)]:
           ba_val, p_ba = ba_state
            # Calculate P(bd|ba)
            if ba val:
               p_bd_y = ba_y_bd_y
            else:
                p_bd_y = ba_n_bd_y
            p_bd_n = 1 - p_bd_y
            for bd_state in [(True, p_bd_y), (False, p_bd_n)]:
                bd val, p bd = bd state
                for no charte in [/Thus 1) /Eales (A)]. # Wolll multiply by actual D/ne) laten
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TOT THE STREET IN [(IT ME, 1), (Taise, 0/]. # WE II MULTIPLY BY ACCUAL F(ITC) TAKEN
                    nc_val, _ = nc_state
                    # Calculate P(bf|bd,nc)
                    if bd_val and nc_val:
                        p_bf_y = bd_y_nc_y_bf_y
                    elif bd_val and not nc_val:
                        p_bf_y = bd_y_nc_n_bf_y
                    elif not bd_val and nc_val:
                        p_bf_y = bd_n_nc_y_bf_y
                    else:
                        p_bf_y = bd_n_nc_n_bf_y
                    p_bf_n = 1 - p_bf_y
                    # Calculate P(-1,-gg|bf)
                    for bf_state in [(True, p_bf_y), (False, p_bf_n)]:
                        bf_val, p_bf = bf_state
                        if bf_val:
                            p_evidence = l_n_bf_y * bf_y_gg_n
                        else:
                            p_evidence = l_n_bf_n * bf_n_gg_n
                        # Add contribution to total probability
                        path_prob = p_fb * p_ba * p_bd * p_bf * p_evidence
                        if nc_val:
                            p_evidence_given_ab_y += path_prob * p_nc_y_given_ab_y
                            p_evidence_given_ab_n += path_prob * p_nc_y_given_ab_n
                            p_evidence_given_ab_y += path_prob * (1 - p_nc_y_given_ab_y)
                            p_evidence_given_ab_n += path_prob * (1 - p_nc_y_given_ab_n)
    \# Calculate P(-1,-gg) using total probability theorem
    {\tt p\_evidence = p\_evidence\_given\_ab\_y * ab\_y + p\_evidence\_given\_ab\_n * ab\_n}
    # Calculate P(+ab|-1,-gg) using Bayes' rule
    p_ab_given_evidence = (p_evidence_given_ab_y * ab_y) / p_evidence
    return p_ab_given_evidence
print("p(+cws|+fb)")
R2 = car_fanbelt(G)
print(f"R2: P(+cws | +fb): {R2}")
R3 = car_battery_age(G)
print(f"R3: p(+cws|+ba){R3}")
R4 = calc_prob_alternator_given_lights_gasgauge(G)
print(f"R4: p(+ab|-1,-gg){R4}")
print(G.nodes())
```



Bayesian Network - Project 3



p(+cws|+fb)
R2: P(+cws | +fb): 0.3467871999999996
R3: p(+cws|+ba)0.32637360000000004
R4: p(+ab|-l,-gg)0.10940802798428095
['battery_age', 'alternator_broken', 'fanbelt_broken', 'battery_dead', 'no_charging_table', 'battery_flat', 'no_oil', 'lights', 'gas_gau