

Homework 2 Advanced Analytics and Metaheuristics

Group 1: Nicholas Jacob

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1. Grapes of Wrath

- (a) Grimes upper bound on grapes used for raisins. We know that 25% of the 8.4×10^6 lb crop was “A” quality with the rest being “B” grade. We also know that “A” grapes average 9 points per pound and “B” grapes average 5 points per pound. Raisins require 8 points or higher in quality. Here we end up with a small LP problem. First we call A the number of pounds of grade “A” grapes and B the number of pounds of grade “B”. Our objective is to maximize the total pounds made into grapes:

$$\text{maximize totalPounds: } A + B;$$

We will have non-negative constraints on A and B , as well as total production constraints

$$A \leq 2100000 \quad \text{and} \quad B \leq 6300000$$

Finally we have the quality constraint:

$$\text{subject to qualityPoints: } 9A + 5B \geq 8(A + B)$$

This one should get a small comment as we want the average number of points to be greater than 8 for our raisins. We code this into our AMPL model and arrive at the following output.

```

Console
AMPL
ampl: model group1_HW2_pla.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 2800000
0 dual simplex iterations (0 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;
A = 2100000
B = 7e+05
totalPounds = 2800000

ampl: |

```

Of course we could have done this with equality but this inequality method worked too.

- (b) To arrive at Bollman's assessment of the price of the grapes, it is important to consider the quality of the inputs. We see that the entire lot was purchased for \$0.28 per pound but some of the grapes are clearly more valuable. With the 25% breakdown of type "A" to "B" we must account for that. Moreover we must use the ratings to arrive at the appropriate value per pound of each type, "A" is rated 9 and "B" is rated 5, the higher the rating the more the value to the company. This creates a system of linear equations with y being price of type "A" and z being price of "B" in cents.

$$\begin{cases} 2100000y + 6300000z = 8400000 \times 28 \\ \frac{y}{9} = \frac{z}{5} \end{cases}$$

We solve this system with substitution and arrive at $y = 42$ and $z = 23\frac{1}{3}$ prices reported are cents per pound. Using these prices we see that jelly can be made entirely from the type "B" grapes and will require 19 pounds per product. Therefore the grape cost for jelly is $19z = \$4.43\frac{1}{3}$. The analysis for the other two products is a bit more complicated. Juice requires a 6 point average so if a_j is the amount of high quality grapes and b_j the amount of low quality grapes, we will need

$$9a_j + 5b_j \geq 6(a_j + b_j)$$

or

$$3a \geq b.$$

So for every 1 pound of high quality grapes, we can use at most 3 pounds of low quality grapes. For purposes of computing the fruit cost, we will compute a minimum cost here but when the grapes are divided later, this price may change. The computed minimum price for the juice grapes is

$$15.5 \frac{(y + 3z)}{4} = \$4.34$$

For the grapes, the minimum rating is 8 so

$$9a_r + 5b_r \geq 8(a_r + b_r)$$

or

$$a_r \geq 3b_r$$

So each pound of low quality grapes we use in raisins, we must just at least 3 pound of high quality grapes. We see then that the minimum price for the input grapes is

$$6.75 \frac{3y + z}{4} = \$2.52.$$

- (c) To formulate for AMPL, we ask for the decision variables to be the amount of each type of grape included in our raisins, jelly and juice. We assume that the cost of the grapes is a fixed ratio based on quality scale points as outlined in 1b but not that the grape input should still be that fixed ratio when computing the price of our input grapes. We consider overhead costs to be fixed and for the accounting department to figure out which product to allocate those costs to based on whatever they do in that department.

Our decision variables will be (a_i, b_i) for $i \in \{\text{raisins}, \text{juice}, \text{jelly}\}$. The objective is to maximize marginal profit (before accounting does their overhead magic). We compute the number of products produced, z_i as

$$z_i = \frac{a_i + b_i}{\text{poundsPerProduct}_i}$$

where $\text{poundsPerProduct}_i$ was given in the table. The fruit cost will be computed as

$$\text{fruitCost}_i = 0.42a_i + 0.233333b_i$$

The our objective is

$$\text{maximize: } \sum_i (sellingPrice_i - variableCost_i) \cdot z_i - fruitCost_i$$

We then have three constraints, quality, demand, and grape availability. For quality, we get that

$$9a_i + 5b_i \geq quality_i(a_i + b_i).$$

Demand constraint is

$$z_i \leq maxDemand_i$$

where the *maxDemand* for raisins was set to infinity. For grape availability, we had

$$\sum_i a_i \leq 2100000 \quad \text{and} \quad \sum_i b_i \leq 6300000.$$

We see the output below highlighting the results.

```
Console
AMPL
ampl: model group1_Hw2_p1c.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 1306119.594
5 dual simplex iterations (3 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;

Weight of Grapes for Each varietyweight :=
jelly   A      0
jelly   B    5510000
juice   A    33750
juice   B   101250
raisins A   2066250
raisins B   688750
;

Number of products produced
The company made 408148.1481481481 units of raisins
The company made 8709.677419354839 units of juice
The company made 290000 units of jelly

Marginal Profit
marginalProfit = 1306120

Left Over Grapes
weightLimit.slack [*] :=
A 0
B 0
;

quality Points Remaining
qualityControl.slack [*] :=
jelly 0
juice 0
raisins 0
;

ampl: |
```

- (d) To discover if we should purchase the additional grade A grapes, we examine two differing solutions. First we run the same model as 1c but include the shadow price of the constraint on the weightLimit.

```

Console
AMPL
;

ampl: model group1_HW2_plda.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 1306119.594
5 dual simplex iterations (3 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;

Shadow Price of Weight Limit
: weightLimit weightLimit.up weightLimit.down :=
A 0.111971 2370000 263333
B 0.13816 8328330 6210000
;

ampl:

```

Some explanation of the output is needed. We see that expanding these weight limits would indeed increase our profit. With type *A* by \$0.11 and *B* by \$0.14 but this is not the price we should be willing to buy more at, rather the price of the current grapes plus this price. So we should be willing to buy more type *A* grapes at up to \$0.53 and type *B* at up to \$0.37. Of course these changes only work with the other fixed and only up to the weightLimit.up displayed in the output. We can take on 270 000 more of type *A* with this same change in the increase in our profits and more than 2 000 000 more pounds of type *B* still creating a profit for any grapes.

Since we decided we could take on more grapes, we also wanted to compute what the optimal purchase would be and what products should be purchased. We expected to only take on the 270 000 pounds of new grapes but that is not what happened. Instead we took them all on and completely changed our production schedule. Making all the raisins we could and putting everything else into jelly, we net an extra ten million before overhead.

```
Console
AMPL
ampl: model group1_HW2_pld.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 1315346.979
5 dual simplex iterations (4 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;
weight :=
jelly   A    0
jelly   B  5500000
juice   A    0
juice   B    0
raisins A  2100000
raisins B   8e+05
;

extraA [*] :=
jelly    0
juice    0
raisins  3e+05
;

weightLimit [*] :=
A  0.0998376
B  0.174561
;

extraGrapes = 0.0198376

ampl:
```

- (e) i. Mr Thomas forgot about the quality constraint and considered type *A* and *B* grapes to be the same cost. We make only raisins in his model and make the largest marginal profit. You would need for type *A* or any type to be just slightly less than \$0.50 per pound to purchase.

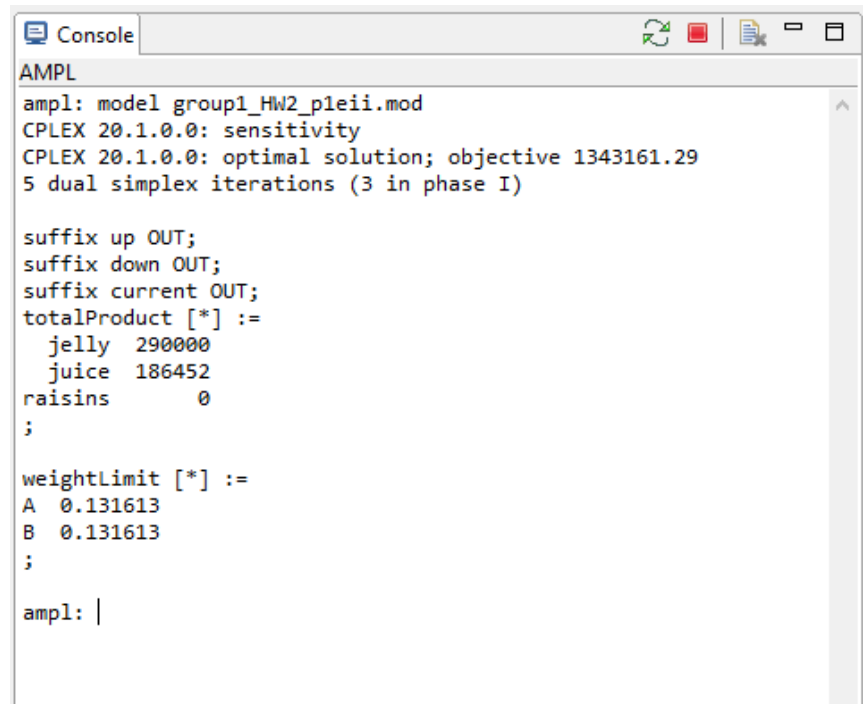
```
Console
AMPL
ampl: model group1_Hw2_p1ei.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 1779555.556
0 dual simplex iterations (0 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;
totalProduct [*] :=
    jelly      0
    juice      0
    raisins 1244440
;

weightLimit [*] :=
A  0.211852
B  0.211852
;

ampl:
```

- ii. Ms Bollman considers the fruit cost to be fixed no matter what the actual input are into each product. So we will use the marginal profit for each product to compute the total marginal profit. She does consider the restrictions on grape quality. This model also suggests that she would be willing to pay up to \$0.55 per pound of type A grapes. The product break down here include all the jelly the market will bear and the rest into juice.



```
AMPL
ampl: model group1_HW2_p1eii.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 1343161.29
5 dual simplex iterations (3 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;
totalProduct [*] :=
  jelly 290000
  juice 186452
  raisins 0
;

weightLimit [*] :=
A 0.131613
B 0.131613
;

ampl: |
```

iii. While both of our teammates report accurate results under their assumptions, they have forgotten important parts of our model. Mr Thomas has ignored quality controls and assumed all grapes at the same price point. Ms Bollman's analysis include accounting for grape quality but when she computes the marginal profit, she does not include the fact that juice made with higher quality grapes will in fact cost our company more. My model deals with all of these. It also has a nice balance of products to keep our customers satisfied. One limitation of our model is the lack of accounting of the overhead costs. We are unclear if those should be attributed to each product in the way that Mr Thomas has done or if the accounting can be done in other ways to minimize our tax liability.

2. Titan's Dealings

- (a) To maximize Titan's return on the investment portfolio, we create our decision variable to be the amount invested in each Project, $amountInvested_a$. We have a max amount that we are allowed

to invest in each

$$amountInvested_a \leq maxAmount_a.$$

Then we have the trickiest part of this exercise, the investment constraint. We start with \$1 000 000. We use the cash-flow table presented in the email from Sharizi, $returns_{y,a}$ indexed on both year and project. For 2021, we ask that the amount remaining, $amountNotInvested_y$,

$$amountNotInvested_{2021} = 1000000 + \sum_{a \in projects} amountInvested_a \cdot returns_{2021,a} \geq 0.$$

This constraint requires that we can only invest up to our \$1 000 000. We should not that all cash-flow in 2021 is negative as we are only investing in that year. For the remaining three years, we have a similar function just without the initial cash but we allow for held over funds to gain interest

$$amountNotInvested_y = 1.06 \cdot amountNotInvested_{y-1} + \sum_{a \in projects} amountInvested_a \cdot returns_{y,a} \geq 0.$$

Some cash-flows are positive and some negative but we do not allow for borrowing of funds. Lastly we succinctly state our objective

$$\text{maximize: } amountNotInvested_{2024}$$

This appears a bit odd as the objective but in year 2024, all returns will come back (no outflows) and this will give us the total returns of all investments.

Some assumptions made here is that there is a 6% return on all funds held over from year to year. No other opportunities for projects present themselves. All funds are withdrawn in 2024. The cash-flow table is true and accurate. Inflation will not degrade the value of our funds. Here is our AMPL output:

```
Console
AMPL
ampl: model group1_Hw2_p2a.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 1797600
4 dual simplex iterations (3 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;
amountInvested [*] :=
A 5e+05
B 0
C 0
D 5e+05
E 659000
;

ampl:
```

- (b) There are two shadow prices within our model. One associated with the constraint on the positivity of the *amountNotInvested* and another on the *maxInvestments*. Both are displayed in the output below.

```

Console
AMPL
ampl: model group1_HW2_p2b.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 1797600
1 dual simplex iterations (0 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;
amountInvested [*] :=
A 5e+05
B 0
C 0
D 5e+05
E 659000
;

: maxInvestments maxInvestments.up maxInvestments.down :=
A 0.0952 0 0
B 0 0 0
C 0 0 0
D 0 0 0
E 0 0 0
;

: maxAvailable maxAvailable.up maxAvailable.down :=
2021 0.17696 1141510 919010
2022 0 1e+20 910000
2023 0.34 1214600 464600
2024 0 1e+20 -606584
;

ampl:

```

We interpret the shadow price of *maxInvestments* as the additional \$0.095 we would earn for every dollar more we could invest in project *A*. So this is an additional way we could earn some more funds by moving that cap. We do however note here that the availability of moving that cap up or down appears to be zero. We think that is due to the fact that all the dollars were put elsewhere in the year that investment in *A* was possible. This shows that simply raising that investment cap will require other changes to the model.

Next we examine the shadow price of the *maxAvailable* variable. The best way to think about this number would be potential gains we could derive by borrowing funds. In the first year, we could

earn 17 cents on any extra dollar included in the model. We see that \$1 000 000 is in-between our up and down values. This is because that was the amount we actually had to invest. We see again there is excellent opportunity to borrow money in 2023 and increase our returns on projects. We assume any funds borrowed in 2023 will be invested in project *E* but the shadow price is slightly less than the actual rate of return on the investment *E*. It would not be prudent to borrow to invest in any other years. We speculate that these rates might be used as a hurdle rate in future years for evaluating any new project.

- (c) We explore several ways to examine this question. First we will look at the opportunity cost for each of our decision variables. This output is presented below by using .current method.

```

AMPL
ampl: model group1_HW2_p2c.mod
CPLEX 20.1.0.0: sensitivity
CPLEX 20.1.0.0: optimal solution; objective 1797600
4 dual simplex iterations (3 in phase I)

suffix up OUT;
suffix down OUT;
suffix current OUT;
amountInvested [*] :=
A 5e+05
B 0
C 0
D 5e+05
E 659000
;

: amountInvested.current amountInvested.up amountInvested.down
A 0.206064 1e+20 0.110864
B 0.1944 0.2584 -1e+20
C 0.044944 0.162544 -1e+20
D 0.558984 0.654184 0.441384
E 0.34 0.440858 0.267769
;

ampl: |

```

- (d) Let's look at the NPV's and IRR's of each of these investments. We see the NPV's (with the hurdle rate at 10%) as follows

$$NPV_F = -1 + \frac{0.8}{1.1} + \frac{0.45}{1.1^2} = 0.099$$

and

$$NPV_G = -1 + \frac{1.1}{1.1} + \frac{0.15}{1.1^3} = 0.112$$

We can also compute the IRR's by setting the NPV to zero and solving for the rate r . For investment F

$$0 = -1 + \frac{0.8}{1+r} + \frac{0.45}{(1+r)^2} = -1 + \frac{1.25 + 0.8r}{(1+r)^2}$$

Simplifying to the quadratic equation

$$r^2 + 1.2r - 0.25 = 0$$

So

$$r_F = \frac{-1.2 \pm \sqrt{1.2^2 - 4 \cdot (-0.25)}}{2} = 18.1\%$$

of course we ignore the negative value. We follow a similar process needing to solve the equation

$$0 = -1 + \frac{1.1}{1+r} + \frac{0.15}{(1+r)^3}.$$

Arrives at

$$0 = r^3 + 1.9r^2 + 0.8r - 0.25$$

(but solved a cubic) and arrive at

$$r_G = 20.4\%$$

(e)