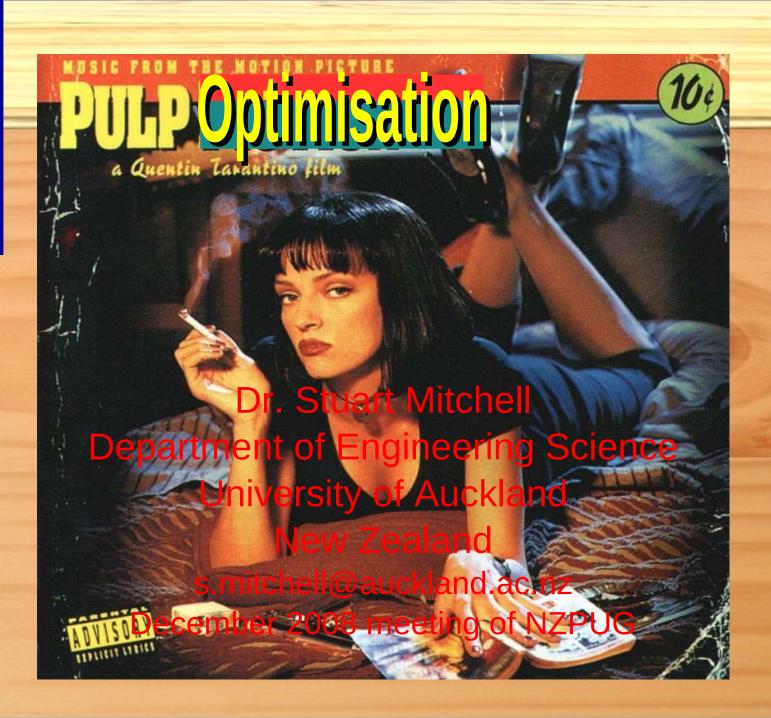


The University of Auckland



# Contents of presentation

What is Mathematical Programing
The Whiskas Problem
PulP
Further PulP examples

# What is Mathematical programing

- A simple mathematically precise way of stating optimisation problems.
- Use mathematically rigorous ways to find a solution
- Examples of Optimisation Problems that can be solved with MP:
  - Shortest Path Problem
  - Scheduling Problems (Set partitioning)
  - Knapsack problems
  - Blending Problems

# The Whiskas blending problem

- Taken from http://130.216.209.237/engsci392/pulp/ABlendingProblem
- Whiskas cat food want to produce their cat food products as cheaply as possible while ensuring they meet the stated nutritional analysis requirements shown on the cans.
- Thus they want to vary the quantities of each ingredient used (the main ingredients being chicken, beef, mutton, rice, wheat and gel) while still meeting their nutritional standards.

## The Whiskas blending problem

NUTRITIONAL ANALYSIS:
Minimum% Crude Protein. 8.0
Minimum% Crude Fat 6.0
Maximum% Crude Fibre 2.0
Max % Salt (Naturally Occurring) 0.4

\$/kg	Protein	Fat	Fibre	Salt
Chicken \$13	0.100	0.080	0.001	0.002
Beef \$8	0.200	0.100	0.005	0.005
Mutton \$10	0.150	0.110	0.003	0.007
Rice \$2	0.000	0.010	0.100	0.002
Wheat bran	0.040	0.010	0.150	0.008
Gel \$1	=	7.	-	

## Whiskas blending problem

- We wish to identify decision variables
- Assume we only have chicken and beef
- Let
  - $-x_c$  = the percentage of chicken meat
  - $-x_{b}$  = the percentage of beef used
- Then we wish to minimise \$13x<sub>c</sub> + \$8x<sub>b</sub>

## Whiskas Blending Problem

- What about the nutritional requirements
- Subject to  $x_c + x_b = 100$   $0.100x_c + 0.200x_b >= 8.0$   $0.080x_c + 0.100x_b >= 6.0$   $0.001x_c + 0.005x_b <= 2.0$  $0.002x_c + 0.005x_b <= 0.4$
- $x_c >= 0, x_b >= 0$

## MP Model

Let:  $I \in \{c, b, m, w, g\}$  the set of ingredients  $x_i$  be the percentage of ingredient i in the cat food  $i \in I$   $C_i$  be the cost of ingredient i  $i \in I$   $P_i$  be the protien content of ingredient i  $i \in I$   $F_i$  be the fat content of ingredient i  $i \in I$   $Fb_i$  be the fibre content of ingredient i  $i \in I$   $S_i$  be the salt content of ingredient i  $i \in I$ 

$$\min \sum_{i \in I} C_i x_i$$
s.t.
$$\sum_{i \in I} x_i = 100$$

$$\sum_{i \in I} F_i x_i \ge 8$$

$$\sum_{i \in I} Fb_i x_i \le 2$$

$$\sum_{i \in I} S_i x_i \le 0.4$$

$$x_i \ge 0 \ \forall i \in I$$

## Ok where is the python

- On google code you can find pulp-or http://code.google.com/p/pulp-or/
- This is a python module that allows the easy statement and solution of linear programing problems.
- Pulp leverages features of python and the open source optimisation libraries Coin-or

# Whiskas model in python

```
The Full Whiskas Model Python Formulation for the PuLP Modeller
Authors: Antony Phillips, Dr Stuart Mitchell 2007
# Import PuLP modeler functions
from pulp import *
# Creates a list of the Ingredients
Ingredients = ['CHICKEN', 'BEEF', 'MUTTON', 'RICE', 'WHEAT', 'GEL']
# A dictionary of the costs of each of the Ingredients is created
costs = {'CHICKEN': 0.013,}
     'BEEF': 0.008,
     'MUTTON': 0.010,
     'RICE': 0.002,
     'WHEAT': 0.005.
     'GEL': 0.001}
# A dictionary of the protein percent in each of the Ingredients is created
proteinPercent = {'CHICKEN': 0.100,
           'BEEF': 0.200,
           'MUTTON': 0.150,
           'RICE': 0.000,
           'WHEAT': 0.040,
           'GEL': 0.000}
# A dictionary of the fat percent in each of the Ingredients is created
fatPercent = {'CHICKEN': 0.080,}
        'BEEF': 0.100,
        'MUTTON': 0.110,
        'RICE': 0.010,
        'WHEAT': 0.010,
        'GEL': 0.000}
# A dictionary of the fibre percent in each of the Ingredients is created
fibrePercent = {'CHICKEN': 0.001,
          'BEEF': 0.005.
          'MUTTON': 0.003.
          'RICE': 0.100.
          'WHEAT': 0.150.
          'CEI '- 0 0001
```

```
# Create the 'prob' variable to contain the problem data
prob = LpProblem("The Whiskas Problem", LpMinimize)
# A dictionary called 'Vars' is created to contain the referenced Variables
vars = LpVariable.dicts("Ingr",Ingredients,0)
# The objective function is added to 'prob' first
prob += lpSum([costs[i]*vars[i] for i in Ingredients]), "Total Cost of Ingredients per can"
# The five constraints are added to 'prob'
prob += lpSum([vars[i] for i in Ingredients]) == 100, "PercentagesSum"
prob += lpSum([proteinPercent[i] * vars[i] for i in Ingredients]) >= 8.0, "ProteinRequirement"
prob += lpSum([fatPercent[i] * vars[i] for i in Ingredients]) >= 6.0, "FatRequirement"
prob += lpSum([fibrePercent[i] * vars[i] for i in Ingredients]) <= 2.0, "FibreRequirement"
prob += lpSum([saltPercent[i] * vars[i] for i in Ingredients]) <= 0.4, "SaltRequirement"
# The problem data is written to an .lp file
prob.writeLP("WhiskasModel2.lp")
# The problem is solved using PuLP's choice of Solver
prob.solve()
# The status of the solution is printed to the screen
print "Status:", LpStatus[prob.status]
# Each of the variables is printed with it's resolved optimum value
for v in prob.variables():
  print v.name, "=", v.varValue
# The optimised objective function value is printed to the screen
print "Total Cost of Ingredients per can = ", value(prob.objective)
```

#### The open source future for OR

- Presently the Computational OR tools used, taught within this department, are closed source.
  - Excel /Storm
  - AMPL, GAMS
  - CPLEX, EXPRESS, ZIP
- Students can not afford commercial licences of this software
- Students cannot see how this software works.

#### The open source future for OR

- Outcomes for students
  - Ability to access free (no cost) software to implement their own solutions once they graduate
  - Ability to access free (open) source code to see how the algorithms are implemented.
    - Imagine the difference to 391??
  - The ability to improve the software they use.

#### PuLP

- PuLP is a python module that allows the easy expression of Mathematical Programs
- PuLP is built to interface with separate solvers
- PuLP is similar in style to:
  - -AMPL
  - -GAMS
  - OPL
  - LINGO
  - FLOPC++ etc.

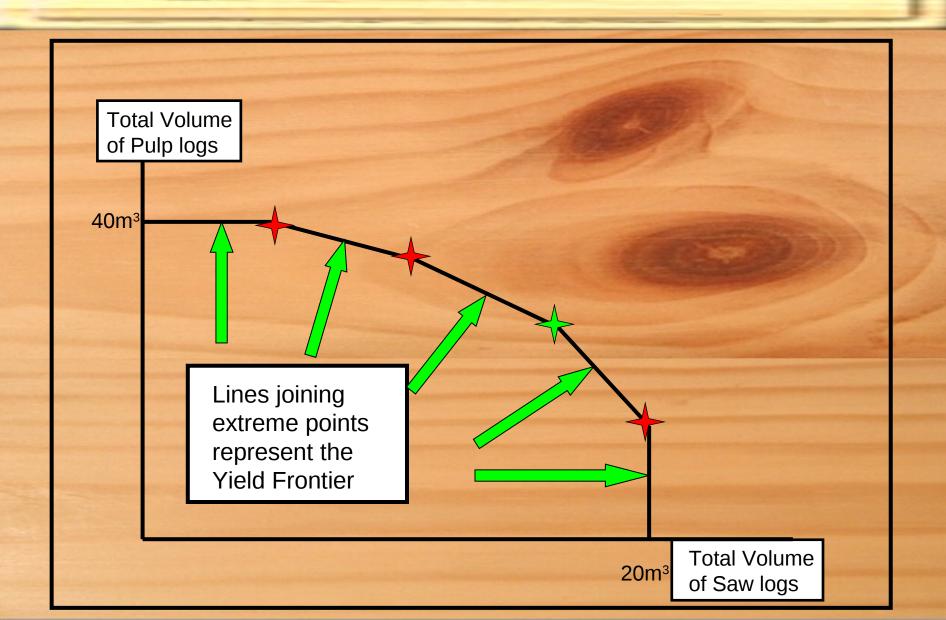
#### PuLP

- Why Python?
  - Core Python syntax leads to the concise statement of MP's
  - Python is a scripting language so no compilation is needed and the code is platform independent
  - Python interfaces easily with external solvers that do the heavy lifting
  - Python comes with 'batteries included'
    - The Python standard library is huge

#### PuLP

- Written initially by J. S. Roy
- Now maintained by S. A. Mitchell
- It is available at http://pulp-or.google-code.com
- Now available for Windows and Linux

# Generating a Yield Frontier



## Generating a Yield Frontier

 Using pulp we formulate the bucking problem (with a single objective) as a set packing problem by log section.

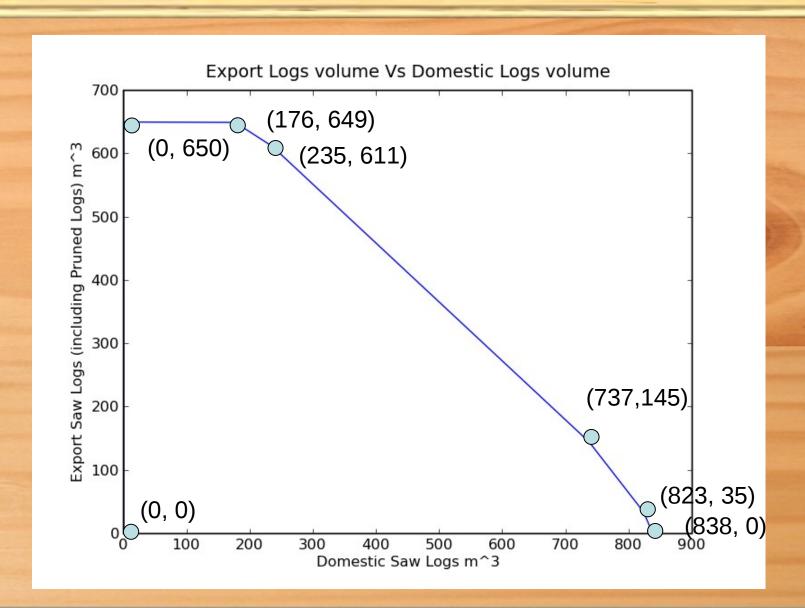
```
lp = LpProblem("Bucking Model", LpMaximize)
#set up the logvolume variables
logvol=LpVariable.dicts("LogVolume(\"%s\")",logtypes,0)
#objective
lp+=lpSum([l.price * logvol[l] for l in logtypes]), "Revenue"
#setup the arc variables
x=LpVariable.dict("x(%s)",f_logs,0,1,LpInteger)
#set up a section set partitioning problem
count = 0
for s in stems:
     slogs = fs_logs[s]
     for i,sec in enumerate(s.sections):
        lp += (lpSum((x[log] for log in slogs)))
                if log.startl <= sec.start
                if log.endl > sec.start)) <= 1
                , "Stem Section(\"%s\",%i)" % (str(s),i))
        count += 1
#add the constraints that link the log volumes
for It in logtypes:
     lp +=( lpSum((log.volume*x[log]
             for log in fl logs[lt]) - logvol[lt] == 0
              , "Logtype_volume(\"%s\")" % str(lt))
```

## Generating a Yield Frontier Using Pulp

- We then iteratively solve the problem to find all extreme supported solutions on the Yield Frontier
- Equivalent to projecting the problem into the log volume space
- I added a module to PuLP that implements projection using Iterative Hull Methods (Lassez, Lassez 1992)

>>> pprob, ppoints = polytope.project(lp, totalvars)

#### Find Yield Frontier for the Dataset



### Find Yield Frontier for a Single Stem

```
\* Total projected *\
Minimize
OBJ: dummy
Subject To
C1: DomSaw + 1.11154598826 ex <= 2669.27592955
C2: DomSaw + 1.34653465347 ex <= 3118.57425743
C3: 1.00863930886 DomSaw + ex <= 2522.60691145
Bounds
dummy = 0
End
```

#### Travelling tournament problem with PuLP

- This problem models the allocation of teams to Home and Away games in a tournament
- A full problem description and datasets are found at Michael Trick's page
- http://mat.tepper.cmu.edu/TOURN/

#### Travelling tournament problem with PuLP

- At IFORS 2008 M. Trick presented an approach to finding lower bounds to this problem using combinatorial benders cuts
- That evening I implemented his algorithm using PuLP
- Along the way I also added Powerset,
   Combination and Permutation operators to PuLP

```
lp = LpProblem("Travelling tournement Master", LpMinimize)
#create variables
triplist = [Trip(t1,p)] for t1 in teams
          for p in
          allpermutations([t for t in
          teams if t = t1, k
          if p[0] \le p[-1]
tripvars = LpVariable.dict("mastervar ",triplist,0,1,LpInteger)
#objective
lp += lpSum([t.cost()*tripvars[t]
              for t in triplist])
#construct constraints to ensure that all teams visit each other
  for t1 in teams:
     for t2 in teams:
        if t1 != t2:
          lp += lpSum([tripvars[t] for t in triplist
                         if t.team == t1
                         if t2 in a.awayteams]) == 1, \
                         "Team %s Visits %s"%(t1,t2)
```

## Further examples

- The 392 course has been converted from AMPL to PuLP http://130.216.209.237/engsci392/pulp/OptimisationWithPuLP
- There you can see a number of different ways to construct problems
- Note that new language features can be added very easily only needing approval from the BDFL