

Problem 1

$$\min_x c^T x \quad \text{s.t.} \quad Ax = b, \quad x \geq 0$$

$$c, x \in \mathbb{R}^n$$

$$b \in \mathbb{R}^m$$

KKT conditions: $L(x, \lambda, s) = c^T x - \lambda^T (Ax - b) - s^T x$

- $\nabla_x L(x^*, \lambda^*, s^*) = c - A^T \lambda^* - s^* = 0$
- $Ax^* = b \quad \hookrightarrow A^T \lambda^* + s^* = c$
- $x^* \geq 0$
- $s^* \geq 0$
- $s^{*T} x^* = 0$

a) The Newton direction may not be defined because $\nabla^2(c^T x) = 0 \neq 0$, i.e. $\nabla^2(c^T x)$ is not positive definite.

b) Definition of convex function

$$f(\alpha x + (1-\alpha)y) \leq \alpha f(x) + (1-\alpha)f(y) \quad \forall \alpha \in [0,1]$$

In LP: $f(x) = c^T x$

• Objective function convex: ✓

$$\begin{aligned} c^T(\alpha x + (1-\alpha)y) &= c^T(\alpha x) + c^T((1-\alpha)y) \\ &= \alpha c^T x + (1-\alpha)c^T y \end{aligned}$$

To have a convex optimization problem, we need

- convex objective function ✓
- linear equality constraints: $Ax = b$ ✓
- concave inequality constraints: ✓
 $x \geq 0 : f(\alpha x + (1-\alpha)y) \geq \alpha f(x) + (1-\alpha)f(y)$
 $(\alpha x + (1-\alpha)y) \geq \alpha x + (1-\alpha)y$

\Rightarrow The linear program is convex.

c) Dual problem: $\max_{\lambda} b^T \lambda \quad \text{s.t.} \quad A^T \lambda \leq c$

Write dual to standard form:

$$\min_{\lambda} -b^T \lambda \quad \text{s.t.} \quad c - A^T \lambda \geq 0$$

$$\mathcal{L}(\lambda, x) = -b^T \lambda - x^T (c - A^T \lambda)$$

KKT conditions:

$$\begin{aligned} \cdot \nabla_{\lambda} \mathcal{L}(\lambda^*, x^*) &= -b + A x^* = 0 \\ &\rightarrow A x^* = b \end{aligned}$$

$$\begin{aligned} \cdot x^* &\geq 0 \\ \cdot s^* &\triangleq (c - A^T \lambda^*) \Rightarrow s^* \geq 0 \\ \cdot s^{*T} x^* &= 0 \end{aligned}$$

$$d) \quad c^T x^* = (A^T \lambda^* + s^*)^T x^* = \lambda^{*T} \underbrace{A x^*}_{=b} + \underbrace{s^{*T} x^*}_{=0}$$

$$= \lambda^{*T} b = b^T \lambda^*$$

$$\Rightarrow \underline{c^T x^* = b^T \lambda^*}$$

e) Basic feasible point:

- The possible point that can be optimal points.
- In LP, this is the vertices of the feasible region.
- $B \subseteq \{1, \dots, n\}$ with exactly m indices s.t. if $i \notin B \Rightarrow x_i = 0$ and $B = [A_i]_{i \in B}$ is a $m \times m$ matrix with columns from A .

f) LICQ: All $\nabla c_i, i \in A(x)$ are linearly independent.

A is full rank \Leftrightarrow all rows of A linearly independent.

This means all constraints are linearly independent, hence ∇c_i are also linearly independent.

Problem 2

$$a) R_I: 2A + 1B \leq 8$$

AB: loading product.

$$R_{II}: 1A + 3B \leq 15$$

$$\rightarrow 2A + 1B + x_3 = 8$$

$$x_3, x_4 \geq 0$$

$$1A + 3B + x_4 = 15$$

\rightarrow slack variables

$$x = \begin{bmatrix} A \\ B \\ x_3 \\ x_4 \end{bmatrix}; \quad A = \begin{bmatrix} 2 & 1 & 1 & 0 \\ 1 & 3 & 0 & 1 \end{bmatrix} \quad b = \begin{bmatrix} 8 \\ 15 \end{bmatrix}$$

$$c^T = [3/2 \quad -1] \quad \text{if price of 1B: } -1$$

$$\Rightarrow \text{LP: } \min_{x} c^T x \quad \text{s.t. } Ax = b, \quad x \geq 0$$

c) Yes, x^* is at the intersection of both constraints, making them both active.

b,d) see figure below.

e) The plot only shows A and B, but we see the iterations move along the constraints $Ax = b$ and $x = 0$. This fits the theory well.

Problem 3

$$\min_x q(x) = \frac{1}{2} x^T G x + x^T c$$

 $G: n \times n, \text{ sym.}$

$$\text{s.t. } a_i^T x = b_i \quad i \in E$$

 $c, x, a_i \in \mathbb{R}^n$

$$a_i^T x \geq b_i \quad i \in I$$

$$a) \quad A(x^*) = E \cup \{i \in I / a_i^T x^* = b_i\}$$

where x^* is the optimal point.

$$b) \quad \mathcal{L}(x^*, \lambda^*) = \frac{1}{2} x^{*T} G x^* + x^{*T} c - \sum_{i \in A(x^*)} \lambda_i (a_i^T x^* - b_i)$$

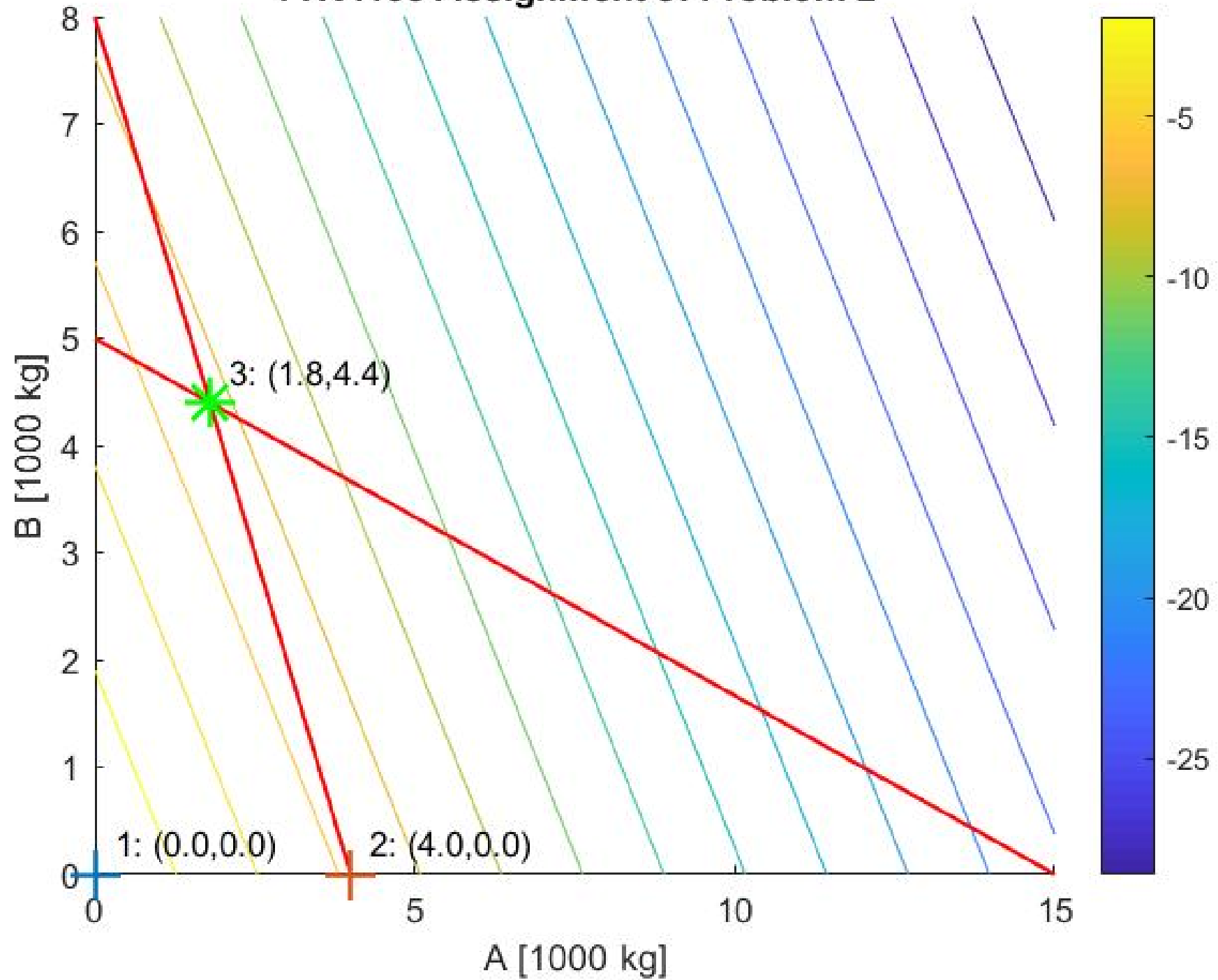
$$\cdot \quad \nabla_x \mathcal{L}(x^*, \lambda^*) = G x^* + c - \sum_{i \in A(x^*)} a_i \lambda_i = 0$$

$$\cdot \quad a_i^T x^* = b_i \quad i \in A(x^*)$$

$$\cdot \quad a_i^T x^* \geq b_i \quad i \in I \setminus A(x^*)$$

$$\cdot \quad \lambda_i^* \geq 0 \quad i \in I \cap A(x^*)$$

TTK4135 Assignment 3: Problem 2



```

1 %% LP definition
2 % x = [A; B; x3; x4];
3 % x3,x4: slack variables
4 Aeq = [2 1 1 0;
5        1 3 0 1];
6 beq = [8; 15];
7 c = [-3/2; -1; 0; 0];
8
9 f = @(a,b) c'*[a; b; 0; 0];
10
11
12 %% Meshgrid and contour
13 [X,Y] = meshgrid(0:1:15,0:1:8);
14 Z = zeros(size(X));
15 % TODO: Is there an easier way to do this?
16 for i = 1:size(X,1)
17     for j = 1:size(X,2)
18         Z(i,j) = f(X(i,j),Y(i,j));
19     end
20 end
21
22 figure(1);
23 hold on;
24 title('TTK4135 Assignment 3: Problem 2');
25 xlabel('A [1000 kg]');
26 ylabel('B [1000 kg]');
27
28 contour(X,Y,Z,15);
29 colorbar;
30
31
32 %% Constraints
33 % see separate file for function: ploteqconstraints
34 ploteqconstraints(Aeq, beq, 'r', 'LineWidth', 1.2);
35
36
37 %% Simplex method
38 x0 = [0 0 beq]';
39 [xopt, fval, iterations] = simplex(c, Aeq, beq, x0, 'report');
40
41
42 %% Plot x* and iterations
43 for i = 1:size(iterations, 2)
44     p = [iterations(1,i), iterations(2,i)];
45     ptxt = sprintf('%d: (%.1f,%.1f)', i, p(1), p(2));
46     xshift = 0.3;
47     yshift = 0.3;
48     text(p(1) + xshift, p(2) + yshift, ptxt);
49
50     if i == size(iterations, 2)
51         plot(p(1), p(2), 'g*', 'MarkerSize', 15, 'LineWidth', 1.5);
52     else
53         plot(p(1), p(2), '+', 'MarkerSize', 15, 'LineWidth', 1.5);
54     end
55 end
56
57

```

```

1 function ploteqconstraints(Aeq, beq, varargin)
2 % Plot equality constraints for a 2D linear optimization problem
3 % in standard form
4 %
5 %   min c'*x    subject to:   A*x = b, x >= 0
6 %       x
7 %
8 %
9 % ploteqconstraints(Aeq, beq, varargin)
10 %
11 %   Aeq: Nx2 matrix
12 %   beq: Nx1 vector
13 %   varargin: Name, Value used for specs for plotting
14
15 N = size(Aeq,1);
16
17 for i = 1:N
18     temp = num2cell(Aeq(i,:));
19     [ai1, ai2] = temp{:};
20     bi = beq(i);
21
22     p1 = [bi / ai1; 0];
23     p2 = [0; bi / ai2];
24
25     x = linspace(p1(1), p2(1));
26     y = linspace(p1(2), p2(2));
27
28     plot(x, y, varargin{:});
29 end % for
30
31 end % function

```