# Multiconvex Optimization in Julia

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#### **Multiconvex functions**

## Definition (Restriction)

For  $f: \mathbf{R}^n \to \mathbf{R}$  and  $\omega \subseteq \{1, \ldots, n\}$ , define the **restriction**  $f_{\omega}(\cdot, \bar{x}): \mathbf{R}^{|\omega|} \to \mathbf{R}$  of f to  $\omega$  to be the function obtained by fixing the coefficients in  $\omega^C$  to their values in  $\bar{x} \in \mathbf{R}^n$ :  $x \mapsto f_{\omega}(x; \bar{x})$ .

## Definition (Multiconvex function)

A function  $f: \mathbf{R}^n \to \mathbf{R}$  is k-convex if there exists a partition  $\Omega = \{\omega_1, \dots, \omega_k\}$  of  $\{1, \dots, n\}$  so that  $f_{\omega_j}$  is convex for every  $j = 1, \dots, k$ .

Multiconvex functions generalize **biconvex** and **multilinear** functions.

- ▶ A 1-convex function is convex; a 2-convex function is biconvex; a 3-convex function is triconvex; etc.
- A multilinear function is multiconvex.

## Multiconvex problems

Consider a (nonconvex) optimization problem

minimize 
$$f_0(x)$$
  
subject to  $f_i(x_{\beta_i}) \leq 0, \quad i = 1, ..., m$   $(\mathcal{P})$ 

with variable  $x \in \mathbf{R}^n$ .

## Definition (Multiconvex problem)

An optimization problem is k-convex if there exists a partition  $\Omega = \{\omega_1, \dots, \omega_k\}$  of  $\{1, \dots, n\}$  with the following properties:

- $f_0$  is k-convex with partition  $\Omega$ ;
- $f_i$  is convex for every  $i = 1, \ldots, m$ ;
- for every constraint  $i=1,\ldots,m$ , there is an element j of the partition with  $\beta_i\subseteq\omega_j$ .

#### MultiConvex.jl

MultiConvex.jl extends Convex.jl to detect and (heuristically) solve multiconvex optimization problems using **disciplined** multiconvex programming:

- simple: less than 300 lines of code
- heuristic solution method: alternating minimization

## Definition (Disciplined multiconvex problem)

A multiconvex optimization problem is a **disciplined** multiconvex problem if

- $f_0$  is k-convex with partition  $\Omega = \{\omega_1, \dots, \omega_k\}$
- $f_0$  restricted to  $\omega_j$  is a disciplined convex function for every  $j=1,\ldots,k$
- $f_i$  is a disciplined convex function for i = 1, ..., m

#### MultiConvex.jl in action

```
# initialize nonconvex problem
n, k = 10, 1
A = rand(n, k) * rand(k, n)
x = Variable(n, k)
y = Variable(k, n)
problem = minimize(sum_squares(A - x*y), x>=0, y>=0)

# perform alternating minimization on the problem
altmin!(problem)
```

### **Conflict graphs**

#### Definition

The **conflict graph** G = (V, E) of a multiconvex expression e is a graph on the variables in the expression:

$$V = \text{variablesin}(e), \qquad E \subseteq V \times V$$

with the property that for any independent set of variables  $\omega$  in the graph, the restriction  $f_{\omega}$  of f to  $\omega$  is convex.

Every multiconvex expression has a (unique) conflict graph.

#### Conflict graphs: recursion

- ► Constant. A constant c is multiconvex with conflict graph  $(\emptyset, \emptyset)$
- ▶ Variable. A variable v is multiconvex with conflict graph  $(v,\emptyset)$
- ► Expressions. The conflict graph of a composite expression is the union of the conflict graphs of its arguments, together with (possibly) a few more edges.
  - multiplication (\*, (x, y)) adds complete bipartite graph on variablesin(x) and variablesin(y)
- Constraints. A constraint is multiconvex iff it is convex.
- ▶ Problems. Problems check their convexity by constructing a certifying partition  $\Omega$  of the conflict graph of the objective that respects the constraints (if one exists).

### **Alternating minimization**

Now that we've found a partition  $\Omega$ , we can use alternating minimization:

(or ADMM, or ...)

#### More information (and code!)

- Convex.jl: http://www.github.com/JuliaOpt/Convex.jl
- MultiConvex.jl: http: //www.github.com/madeleineudell/MultiConvex.jl
- Convex.jl paper: http://arxiv.org/abs/1410.4821