

**Optimizers everywhere -**

# **Optimization in Financial Applications with MATLAB**

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**MathWorks Consulting Services**

# Optimization

Definition of *optimize* in English:

**optimize** 



- 1 Make the best or most effective use of (a situation or resource):

## Optimum (disambiguation)

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From Wikipedia, the free encyclopedia

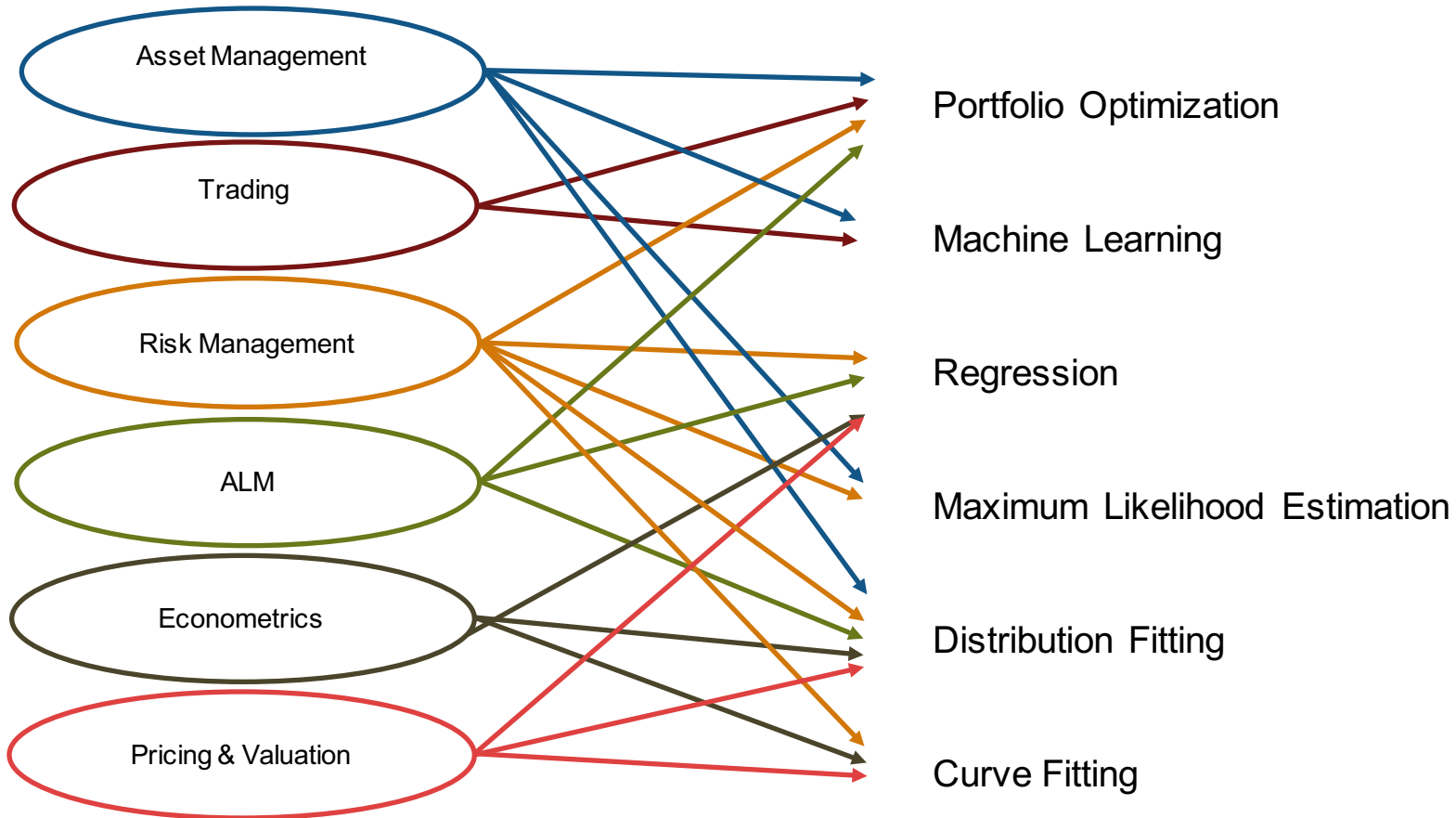
The **optimum** is the best or most favorable condition, or the greatest amount or degree possible under specific sets of comparable circumstances.

# Optimization in Financial Applications with MATLAB

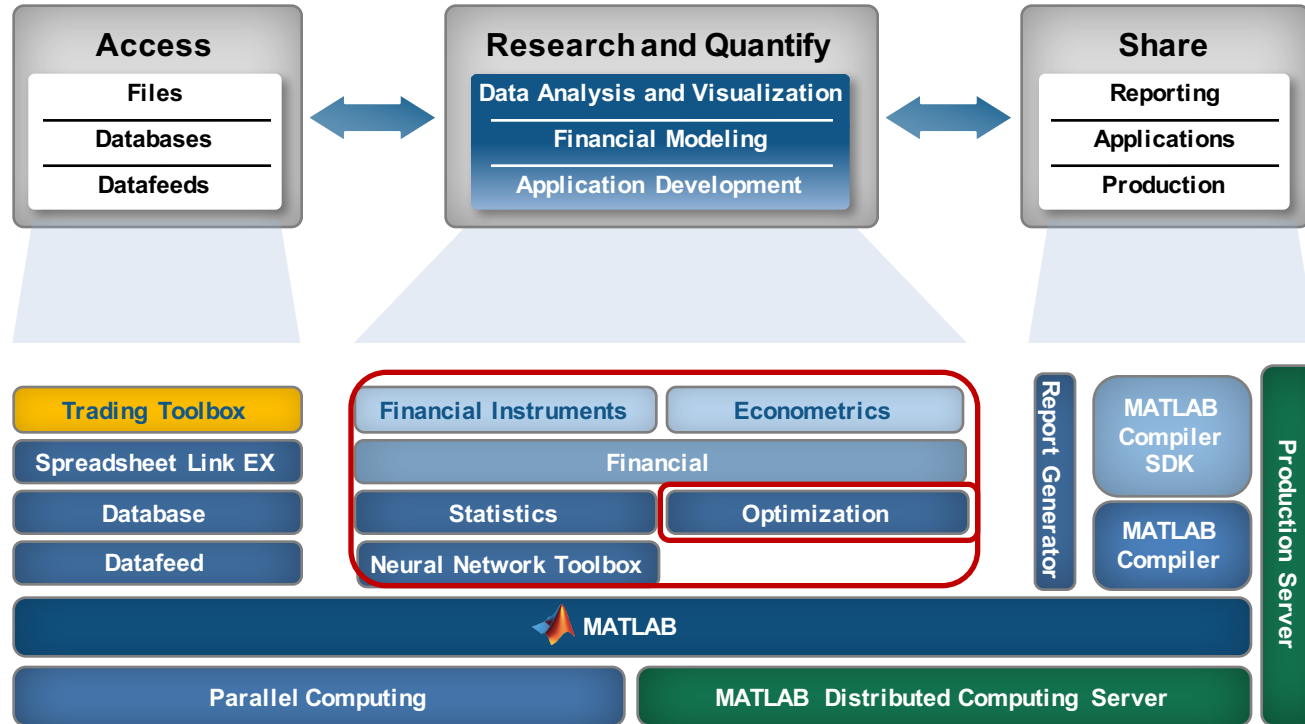
- Financial Optimization
- Optimization Methods
- Customized optimization models

# Financial Optimization

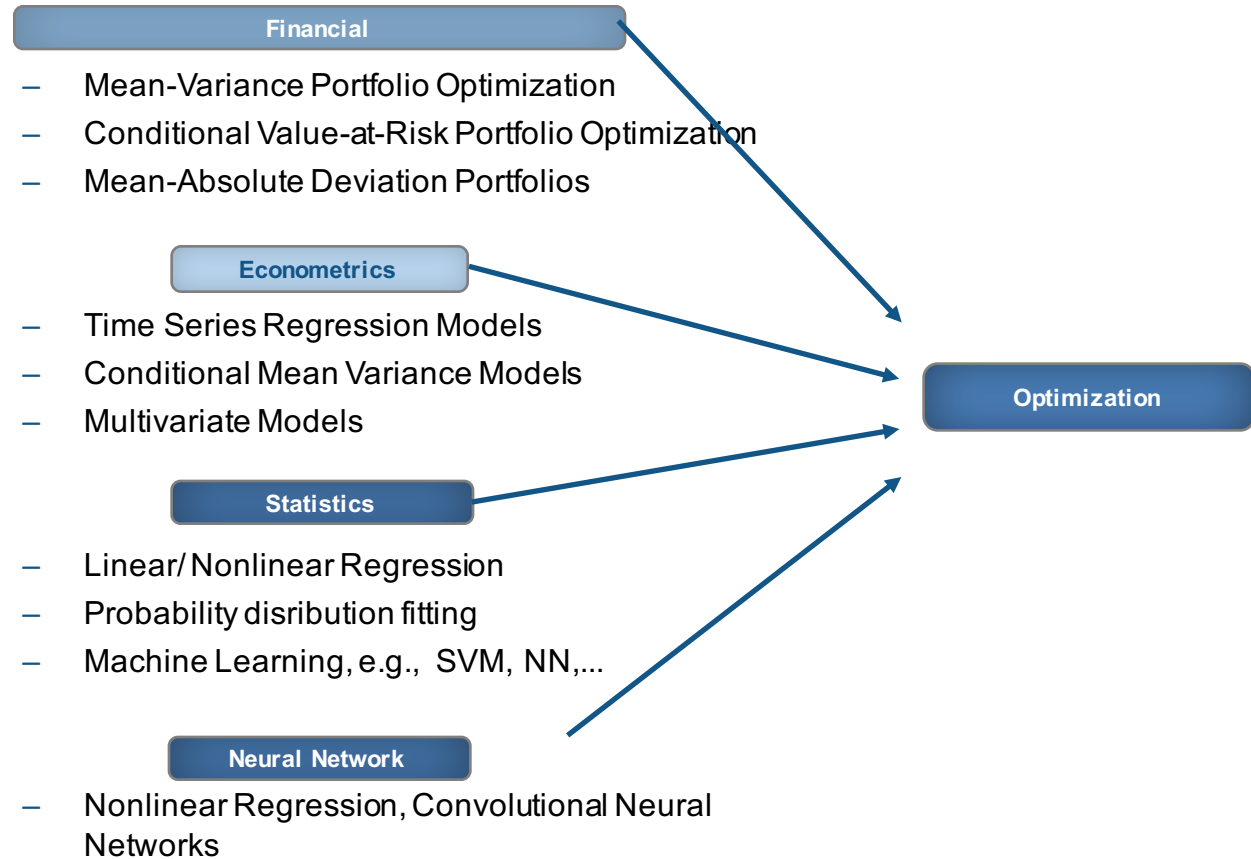
# Financial Applications and Optimization



# MATLAB – The Financial Development Platform



# Financial Optimization within MATLAB



# Optimization Methods



# Optimization Problem

## Objective Function

$$\min_x f(x)$$

Typically a linear or nonlinear function

Decision variables (can be discrete or integer)

## Subject to Constraints

$$Ax \leq b$$

$$A_{eq}x = b_{eq}$$

$$l \leq x \leq u$$

$$c(x) \leq 0$$

$$c_{eq}(x) = 0$$

### Linear constraints

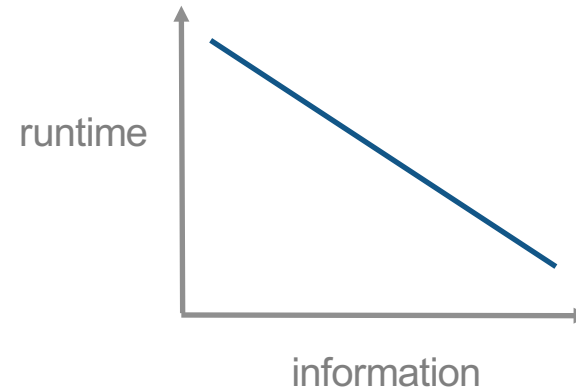
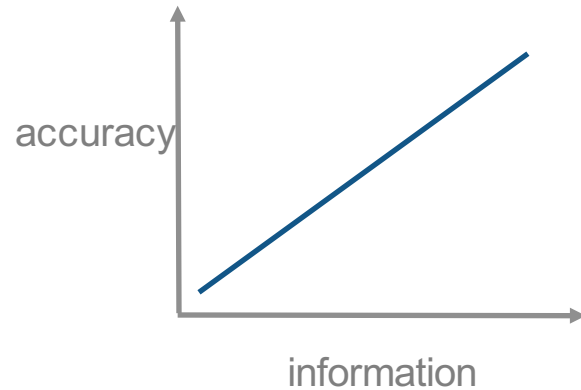
- ⌘ inequalities
- ⌘ equalities
- ⌘ bounds

### Nonlinear constraints

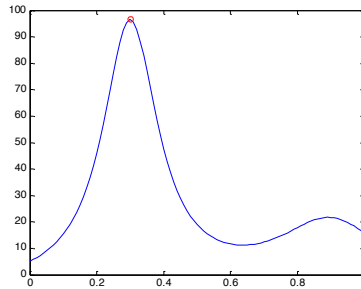
- ⌘ inequalities
- ⌘ equalities

## How to solve an optimization problem ?

What do you know about your optimization problem ?



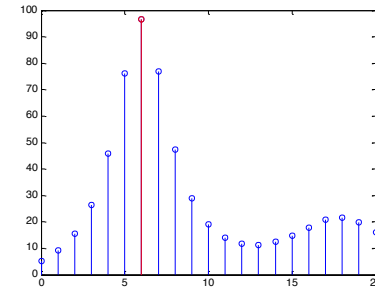
# Variables & Constraints



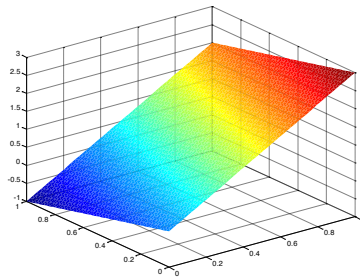
**Variables**

←
→

**Continuous**
**Discrete**



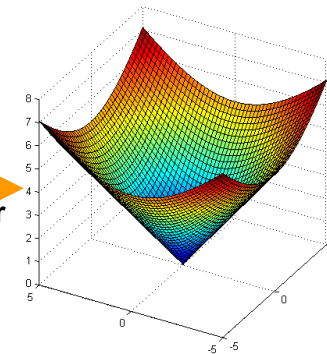
**Integer Programming**



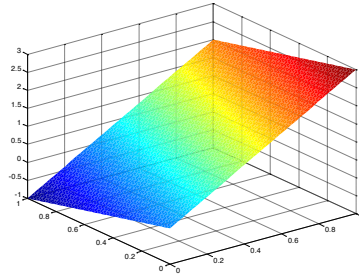
**Constraints**

←
→

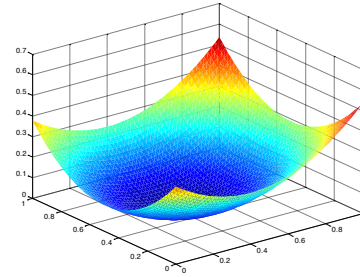
**Linear**
**Nonlinear**



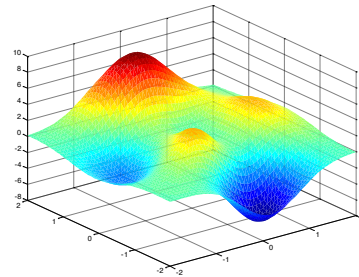
# Objective Function



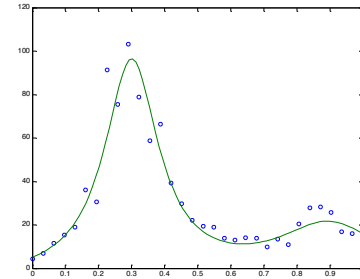
Linear  $f(x) = A^T x$



Quadratic  $f(x) = x^T A x$



General  $f(x)$

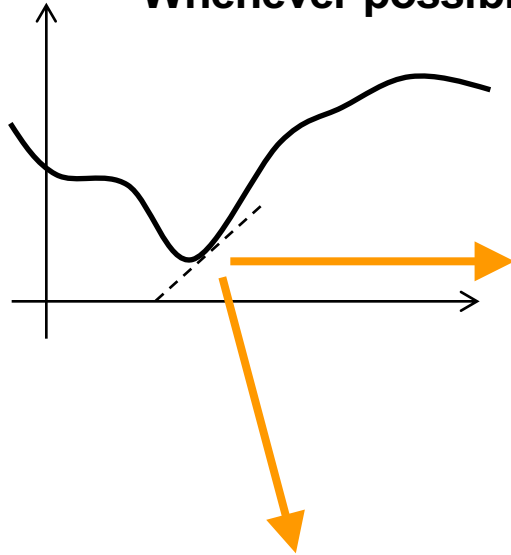


Least-squares/curve fitting

$$f(c) = \sum [g(x_k; c) - y_k]^2$$

# Numerical optimization

Whenever possible, provide gradient/hessian information!

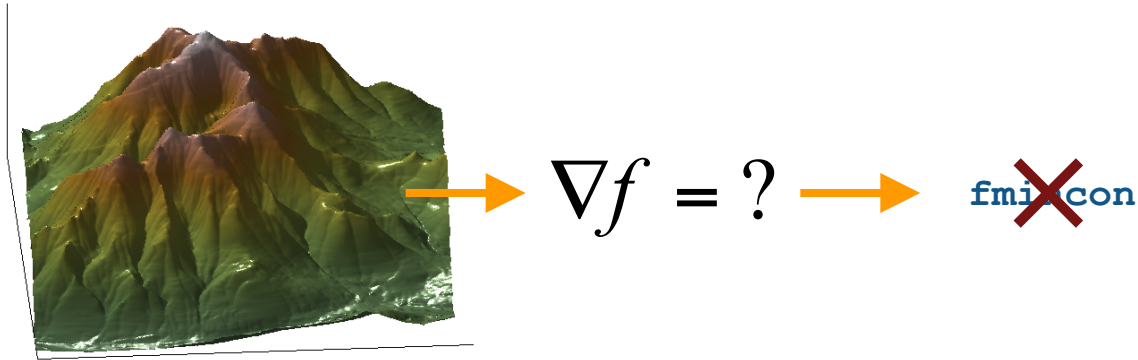


$$f'(x) \approx \frac{f(x + \Delta x) - f(x - \Delta x)}{2\Delta x}$$

```
function [f,df] = objective(x)
f = ... % function value
df = ... % gradient vector
```

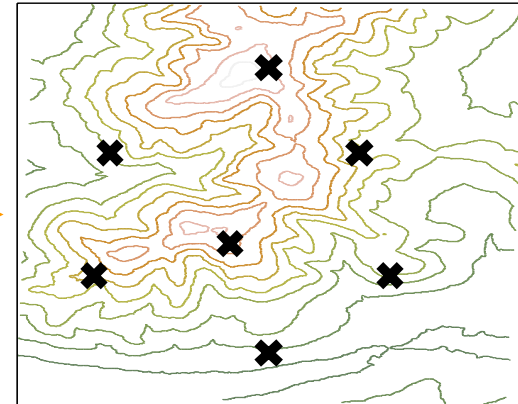
- ✓ Fewer function evaluations
- ✓ More accurate

# Derivative-Free Optimization



Repeatedly sample  
several points

*Direct Search*  
*Genetic Algorithm*



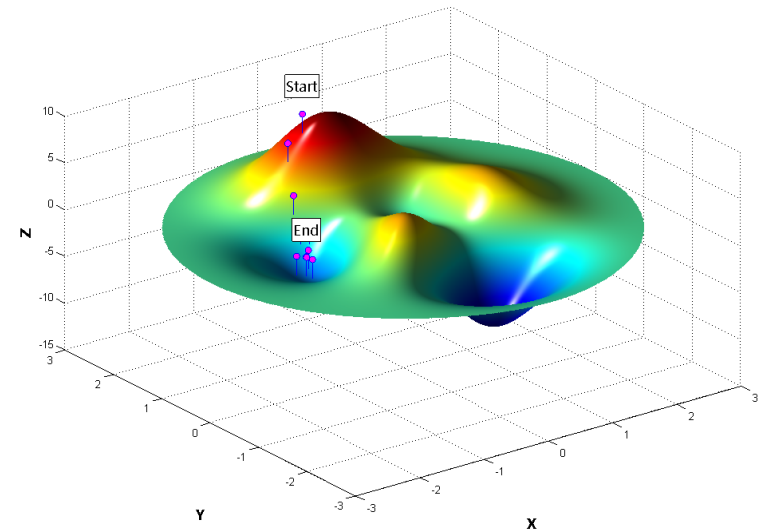
# Approaches in MATLAB

## Local Optimization

- Finds local minima/maxima
- 🔗 Uses supplied gradients or estimates them
- 🔗 Applicable for large scale problems with smooth objective function
- Faster/fewer function evaluations

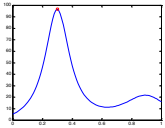
## Global Optimization

- No gradient information required
- Solve problems with non-smooth, discontinuous objective function

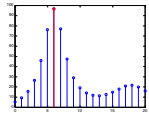


# Solvers

## Variables

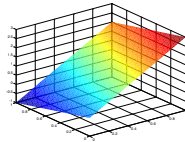


Continuous

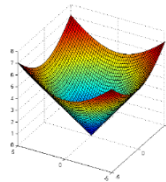


Discrete

## Constraints

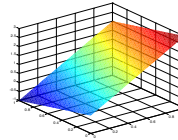


Linear

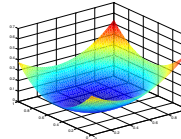


Nonlinear

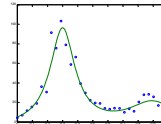
## Objective function



Linear  $f(x) = A^T x$

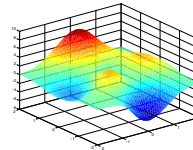


Quadratic  $f(x) = x^T A x$



Least-squares

$$f(c) = \sum [g(x_k; c) - y_k]^2$$



General  $f(x)$

## Solver

`linprog`

Optimization Toolbox

Documentation

CONTENTS

Choosing the Algorithm

- `fmincon` Algorithms
- `fsolve` Algorithms
- `fminunc` Algorithms
- Least Squares Algorithms
- Linear Programming Algorithms
- Quadratic Programming Algorithms
- Large-Scale vs. Medium-Scale Algorithms
- Potential Inaccuracy with Interior-Point Algorithms

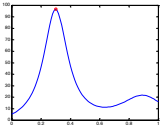
`fminunc`

`fmincon`

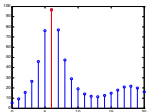


# Solvers

## Variables

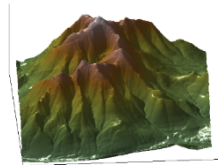


Continuous



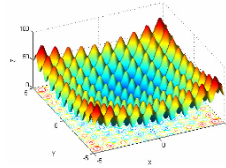
Discrete

## Constraints



Non-smooth

## Objective function



Nonlinear

## Solver

### Global Optimization Toolbox

#### Documentation

#### CONTENTS

### Choosing a Solver

#### Table for Choosing a Solver

There are seven Global Optimization Toolbox solvers:

- `ga` (genetic algorithm)
- `GlobalSearch`
- `MultiStart`
- `patternsearch`, also called direct search
- `particleswarm`
- `simulannealbnd` (simulated annealing)
- `gamultiobj`, which is not a minimizer; see Multiobjective Optimization

## Customized Optimization Models

# Supported Portfolio Optimization Models

Financial Toolbox

- **Mean-Variance Portfolio Optimization**
- **Conditional Value-at-Risk Portfolio Optimization**
- **Mean-Absolute Deviation Portfolio Optimization**

## Customized Objective - Smart Beta

# Customized Portfolio Optimization - Smart Beta

## Smart Beta Example – Risk Parity:

Generate a portfolio where each asset's marginal contribution to risk is equal

- Marginal Contributions for N assets

$$MC_i(x) = x_i \times \frac{\sum_{j=1}^N x_j \text{Cov}(R_i, R_j)}{\sigma}$$

- Minimize distance between all contributions

$$f(x) = \sum_{i=1}^N \sum_{j \neq i} |MC_i(x) - MC_j(x)|$$

# Customized Portfolio Optimization - Smart Beta

- Create marginal risk contributions

$$MC_i(x) = x_i \times \frac{\sum_{j=1}^N x_j \text{Cov}(R_i, R_j)}{\sigma}$$

- Create a distance matrix
- Minimize the total distance between all the contributions

$$f(x) = \sum_{i=1}^N \sum_{j \neq i} |MC_i(x) - MC_j(x)|$$

```
function out = riskCostFunction(wts , covMat)
% Cost function to get equal risk-contribution

% Build up all the marginal risk contributions
rMarg = zeros(size(wts));
for i = 1:length(rMarg)
    rMarg(i) = wts(i) * covMat(i,:) * wts';
end

% Build a distance matrix
[xx,yy] = meshgrid(rMarg);

% Compute total distance between entries
out = sum(sum(abs(xx-yy)));
```

# Customized Portfolio Optimization - Smart Beta

## Smart Beta Example – Risk Parity:

$$\min_x f(x) := \sum_{i=1}^N \sum_{j \neq i} |MC_i(x) - MC_j(x)|$$

$$\text{s.t.} \quad \begin{aligned} x^T \mathbf{e} &= 1 \\ \text{lb} &\leq x \leq \text{ub} \end{aligned}$$

→ Nonlinear objective function with linear equality and bound constraints

# Customized Portfolio Optimization - Smart Beta

- Solve with `fmincon` from MATLAB® Optimization Toolbox™

```
[fixWeights, fVal] = fmincon(@ (x) riskCostFunction(x , covRet) , ...  
    x0 , [] , [] , ...  
    ones(1 , nAsset) , 1 , ...  
    LB , ...  
    UB , [] , o);
```

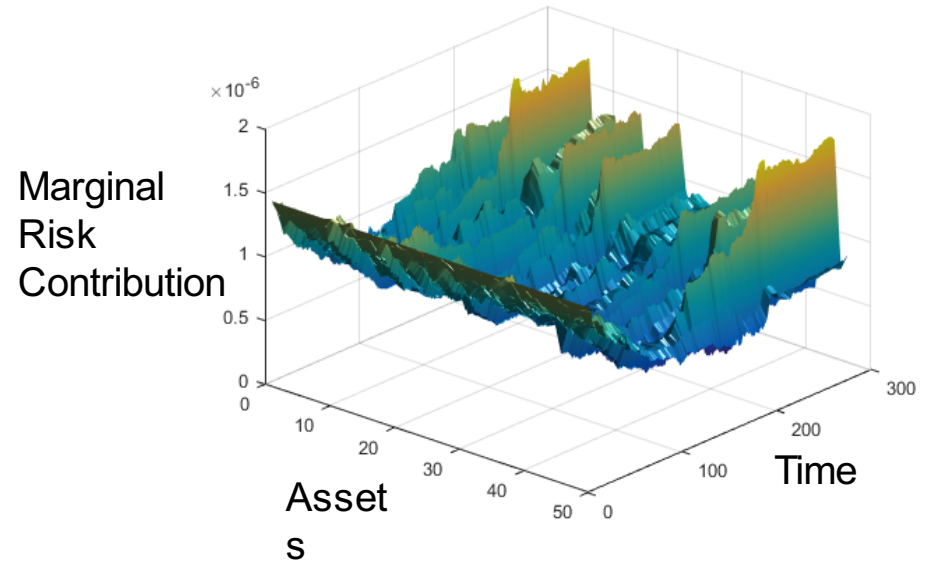
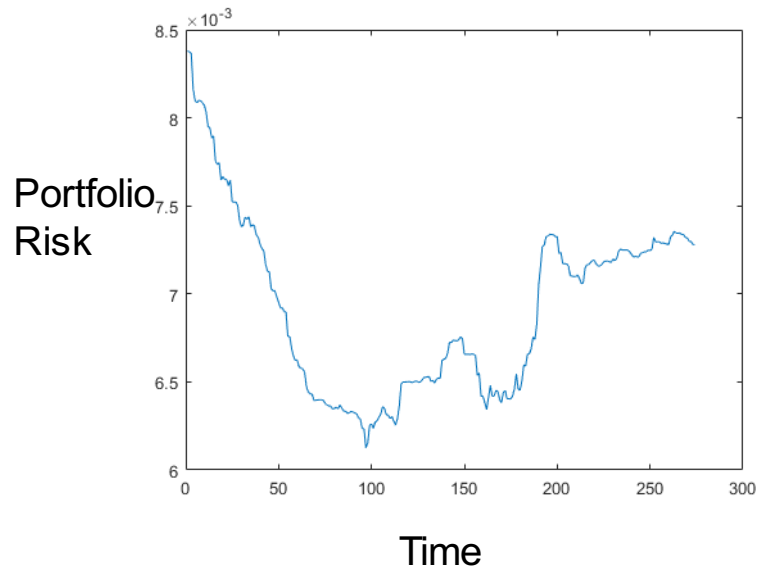
total distance between marginal risk contributions

budget constraint

bounds



# Customized Portfolio Optimization - Smart Beta



## Customized Constraints - Robust Constraints

# Customized Portfolio Optimization - Robust Constraints

## Mean-Variance Portfolio Optimization

$$\begin{aligned}
 & \min x^T C x \\
 & \text{s. t. } m^T x \geq R \\
 & \quad x^T e = 1 \quad \text{budget constraint} \\
 & \quad x \geq 0 \quad \text{lower bound}
 \end{aligned}$$

Covariance matrix  
mean of asset returns

Financial Toolbox

```

m = [ 0.05; 0.1; 0.12; 0.18 ];
C = [ 0.0064 0.00408 0.00192 0;
      0.00408 0.0289 0.0204 0.0119;
      0.00192 0.0204 0.0576 0.0336;
      0 0.0119 0.0336 0.1225 ];

p = Portfolio('assetmean', m, 'assetcovar', C, ...
             'lowerbudget', 1, 'upperbudget', 1, 'lowerbound', 0);

```

# Customized Portfolio Optimization - Robust Constraints

Additional constraints supported by

Financial Toolbox

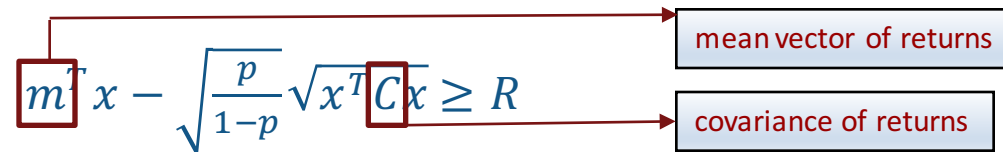
- Linear inequality constraints
- Linear equality constraints
- Group constraints
- Group ratio constraints
- Average turnover constraints
- One-way turnover constraints

# Customized Portfolio Optimization- Robust Constraints

Extend standard model by individual constraints, e.g. robust constraints



→ Deterministic approximation (Chebychev's inequality)



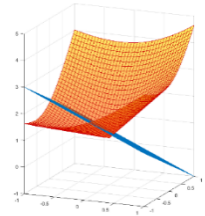
*P. Bonami, M.A. Lejeune, 'An Exact Solution Approach for Portfolio Optimization Problems Under Stochastic and Integer Constraints', Operations Research 2009, Vol. 57, Issue 3*

# Customized Portfolio Optimization - Robust Constraints

Mean-Variance Portfolio Optimization with robust constraint

$$\begin{aligned} \min_x \quad & f(x) := x^T C x \\ \text{s. t.} \quad & x^T e = 1 && \text{budget constraint} \\ & x \geq 0 && \text{lower bound} \end{aligned}$$

$$c(x) := R - m^T x + \sqrt{\frac{p}{1-p}} \sqrt{x^T C x} \leq 0 \quad \text{robust constraint}$$



→ Robust mean variance model is a nonlinear, convex NLP

# Customized Portfolio Optimization- Robust Constraints

- Solve with `fmincon` from MATLAB® Optimization Toolbox™

```
[x, fval] = fmincon(@ (x) quadObj (x, C), ...
    x0, [], [], ...
    ones (1, n), 1, ...
    zeros (n, 1), inf (n, 1), ...
    @(x) probConstr (x, p, m, C, R), ...
    options);
```

Annotations:

- `@ (x) quadObj (x, C)`: mean-variance risk proxy
- `x0, [], []`: start vector
- `ones (1, n), 1`: budget constraint
- `zeros (n, 1), inf (n, 1)`: lower and upper bounds
- `@ (x) probConstr (x, p, m, C, R)`: robust constraint

# Customized Portfolio Optimization- Robust Constraints

Provide gradients

```
options = optimoptions(@fmincon, ...
    'SpecifyObjectiveGradient', true, ...
    'SpecifyConstraintGradient', true, ...
```

```
function [f,gradf] = quadObj(x,C)
    f=x'*C*x; →  $f(x)$ 
    gradf = 2*C*x; →  $\nabla f(x)$ 
end
```

```
function [c,ceq, gradC,gradCeq ] = probConstr(x,p,m,C,R)
    denom = sqrt(x'*C*x);
    c = R - m'*x + p*denom; →  $c(x)$ 
    gradC = -m + p/denom*C*x; →  $\nabla c(x)$ 
    ceq = [];
    gradCeq = [];
```

end



# Customized Portfolio Optimization- Robust Constraints

Provide Hessian

```
options = optimoptions(@fmincon,...
    'SpecifyObjectiveGradient',true,...
    'SpecifyConstraintGradient',true, ...
    'HessianFcn',@(x,lambda) getHessianRobustMV(x,lambda,C,p));
```

$$\nabla^2 f(x) + \sum \nabla^2 c_i(x) \cdot \lambda_i + \sum \nabla^2 ceq_j(x) \cdot \lambda_j$$

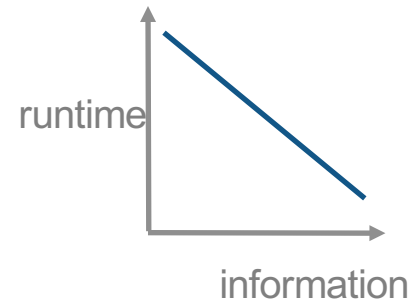
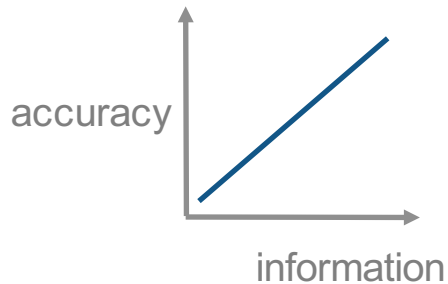
```
function H = getHessianRobustMV(x,lambda,C,p)
% This is the Hessian of
% objective function: quadObj
% constraint function: probConstr

sw = C*x;
dC = (p/sqrt(x'*C*x)).*(C-(sw*sw')./(x'*C*x));
dF = 2*C;
H = dF + lambda.ineqnonlin(1)*dC;

end
```

# Customized Portfolio Optimization - Robust Constraints

assets	result_no_gradients	time_no_gradients	result_with_gradients	time_with_gradients
50	'optimal'	0.1836	'optimal'	0.060784
100	'feasible'	0.56741	'optimal'	0.091847
500	'feasible'	8.5853	'optimal'	2.4931



# Customized Portfolio Optimization - Robust Constraints

Without gradients

Iter	f-count	f(x)	Feasibility	optimality
0	51	6.098570e+01	2.400e+01	1.794e+02
1	102	1.881289e+01	1.757e+01	2.404e+02
2	153	8.850099e+00	1.172e+01	1.583e+02
3	204	2.926182e+00	5.126e+00	5.186e+01
4	255	1.599701e+00	2.437e+00	3.379e+00
5	306	1.269421e+00	5.720e-01	4.025e+00
6	358	1.249948e+00	3.610e-01	2.123e+00
7	409	1.279036e+00	1.805e-03	1.133e-01
8	460	1.042756e+00	9.026e-06	1.134e+00
9	511	5.778849e-01	6.602e-03	5.587e+00
10	562	4.839580e-01	9.742e-04	3.333e+00
11	613	4.824363e-01	9.517e-04	1.264e-01
12	664	3.615160e-01	4.711e-03	9.752e-01
13	715	3.328996e-01	1.925e-04	5.851e-02
14	766	3.204592e-01	6.739e-05	9.337e-02
15	817	3.150931e-01	1.709e-05	3.187e-02
16	868	3.113085e-01	4.433e-12	3.139e-02
17	919	3.108302e-01	1.244e-12	8.114e-03
18	970	3.108004e-01	8.873e-13	2.088e-04
19	1021	3.063753e-01	2.640e-13	5.799e-02
20	1072	3.031614e-01	2.625e-05	7.867e-03
21	1123	3.028625e-01	4.135e-06	4.636e-03
22	1174	3.028082e-01	5.297e-06	2.663e-03
23	1225	3.028131e-01	2.331e-15	3.900e-05
24	1276	3.028131e-01	2.554e-15	3.637e-06

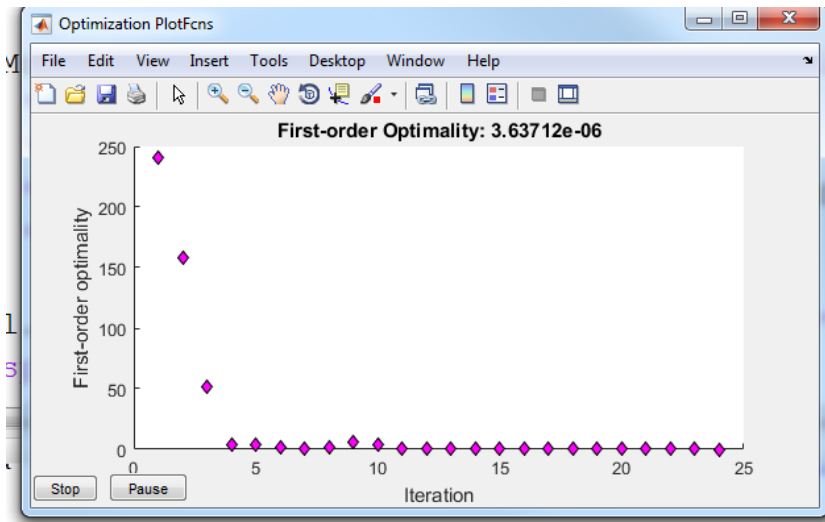
With gradients

Iter	f-count	f(x)	Feasibility	optimality
0	1	6.098570e+01	2.400e+01	1.794e+02
1	2	2.322288e+01	1.430e+01	1.269e+02
2	3	5.419761e+00	5.911e+00	2.018e+01
3	4	1.732759e+00	2.237e+00	5.623e+00
4	6	1.413489e+00	1.431e+00	2.503e+00
5	7	1.217778e+00	1.372e-01	1.418e-01
6	8	8.718262e-01	6.861e-04	1.491e+00
7	9	5.448938e-01	5.936e-03	3.887e+00
8	10	4.831605e-01	4.820e-04	4.325e+00
9	11	3.815757e-01	3.713e-03	1.552e+00
10	12	3.480601e-01	4.921e-04	2.120e+00
11	13	3.301950e-01	1.063e-10	1.102e+00
12	14	3.211171e-01	5.814e-11	2.573e-02
13	15	3.117667e-01	9.545e-05	3.841e-03
14	16	3.108083e-01	1.701e-12	2.069e-04
15	17	3.061063e-01	2.869e-13	6.978e-03
16	18	3.046152e-01	3.464e-12	9.367e-05
17	19	3.029598e-01	2.707e-06	1.094e-02
18	20	3.028149e-01	1.588e-14	1.571e-06

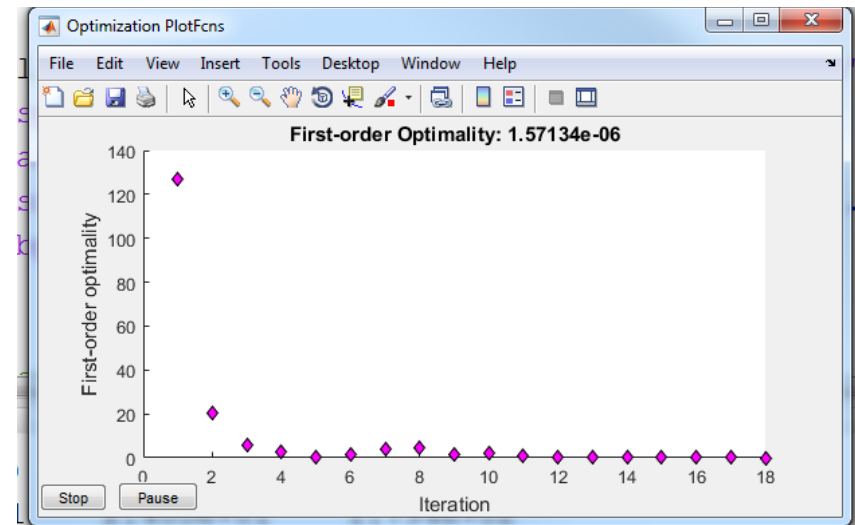
# Customized Portfolio Optimization - Robust Constraints

```
'PlotFcn', @optimplotfirstorderopt
```

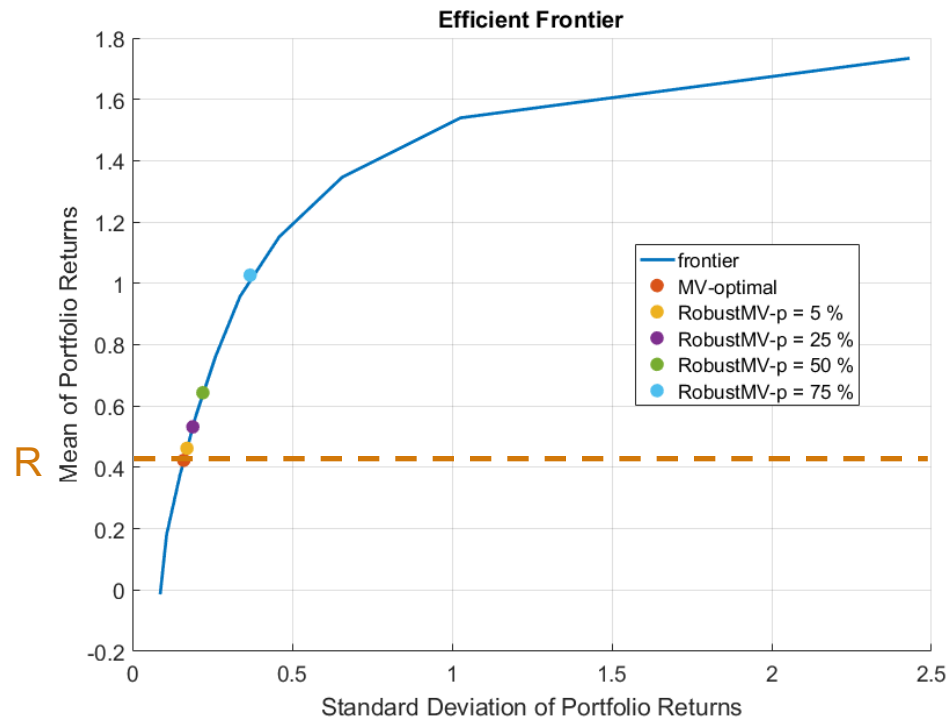
Without gradients



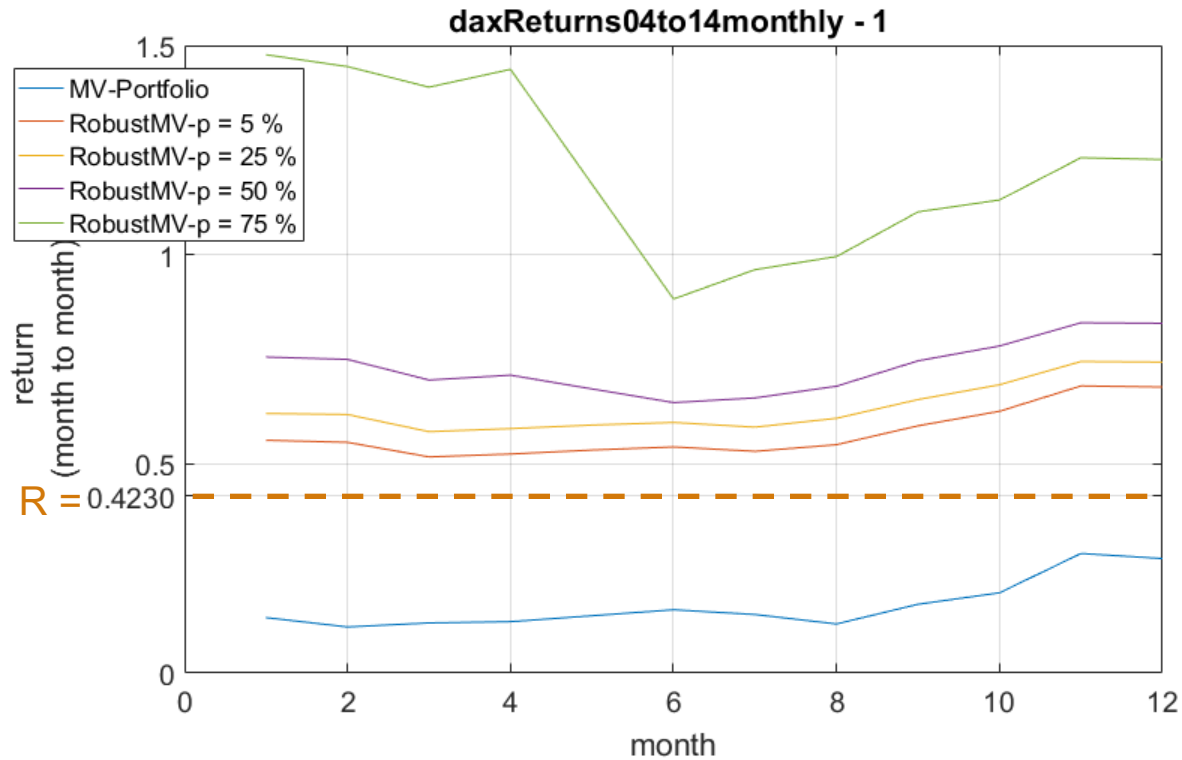
With gradients



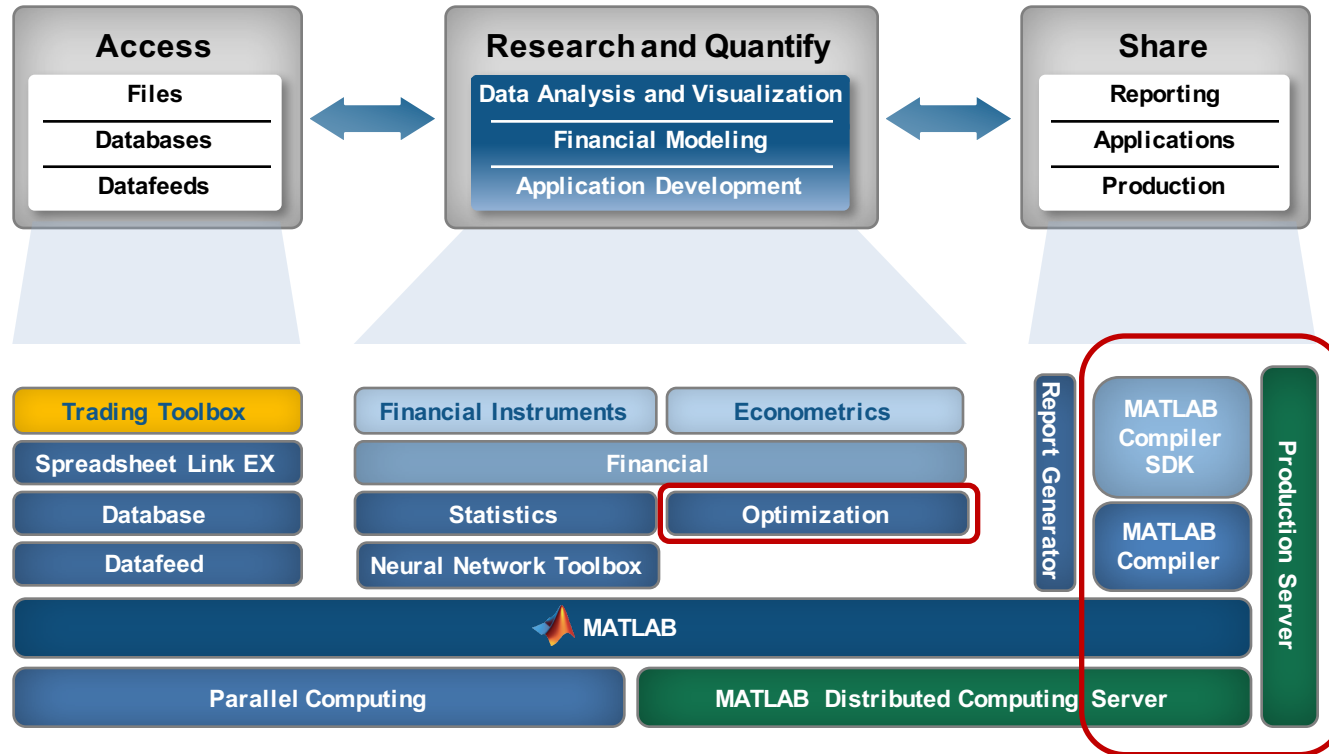
# Customized Portfolio Optimization - Robust Constraints



# Customized Portfolio Optimization



# MATLAB – The Financial Development Platform

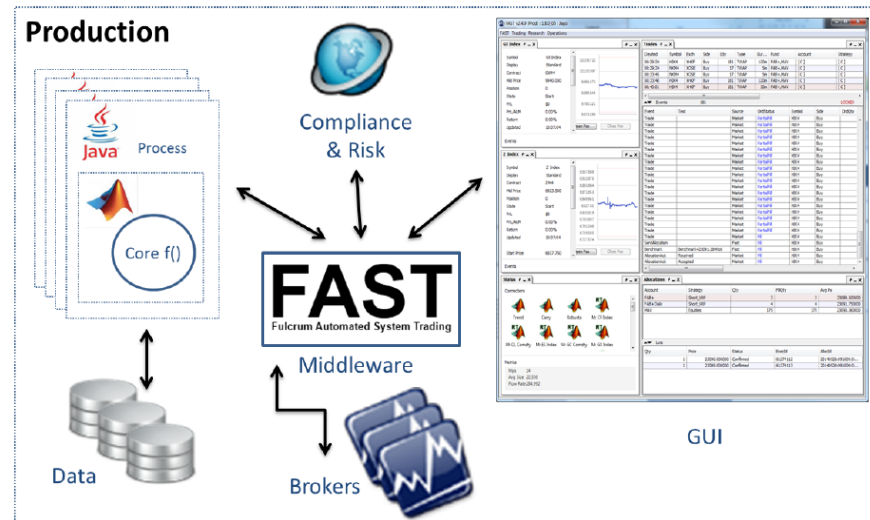


# Customized Portfolio Optimization - Deployment

- Compile your MATLAB optimization model for your dedicated platform
- Make it available for your enterprise environment

## Implementation

Case Study: Production Environment





## Summary

- Optimization for financial applications is built within MATLAB toolboxes covering many standard applications
- A large variety of optimization algorithms available in MATLAB® Optimization Toolbox™ and Global Optimization Toolbox™
- Customized optimization models made easy by quick modeling, advanced optimization process diagnostics and rapid deployment.
- Enhance optimization performance and accuracy by adding maximal information.

**Thank you !**