

# A comparative analysis of various Trust Region subproblems

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# Aim

**Implementation of Trust region subproblem solution variants to arrive at a minima of a given function.**

- Cauchy point method
- Dogleg method
- Conjugate Gradient Steihaug method

**Comparative analysis of the trust region subproblem variants.**

- Scaling factor
- Run time and iterations
- Convergence rates

**Objective functions used:**

- Rosenbrock
- Booth
- Ackley

# Standard Algorithm

**Set** the starting point at  $x_0$ , set the iteration number  $k = 1$

**for**  $k = 1, 2, \dots$

Get the improving step by solving trust-region sub-problem ()

Evaluate  $\rho_k$  from equation()

**if**  $\rho_k < \eta_2$

$$\Delta_{k+1} = t_1 \Delta_k$$

**else**

**if**  $\rho_k > \eta_3$  and  $p_k = \|\Delta_k\|$  (full step and model is a good approximation)

$$\Delta_{k+1} = \min(t_2 \Delta_k, \Delta_M)$$

**else**

$$\Delta_{k+1} = \Delta_k$$

**if**  $\rho_k > \eta_1$

$$x_{k+1} = x_k + p_k$$

**else**

$x_{k+1} = x_k$  (the model is not a good approximation and need to solve another trust-region subproblem within a smaller trust-region)

**end** >

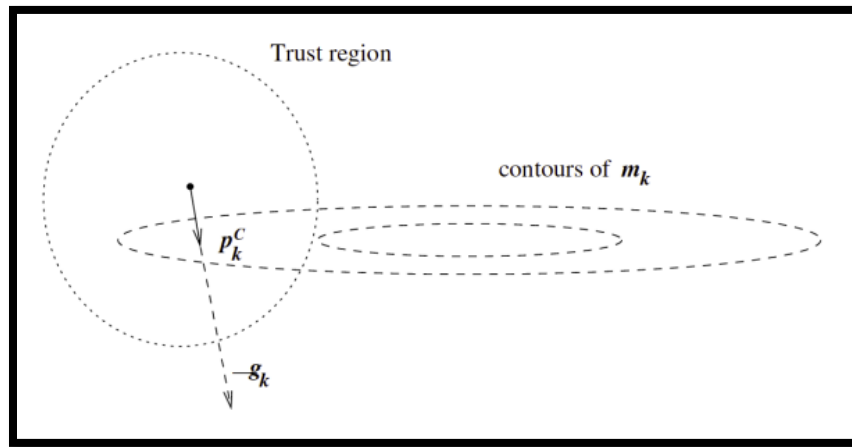
$$\rho_k = \frac{f(x_k) - f(x_k + p_k)}{m_k(0) - m_k(p_k)}$$

# Cauchy point method

$$p_k^C = -\tau_k \frac{\Delta_k}{\|g_k\|} g_k$$

$$\text{if } g_k^T B_k g_k \leq 0 \quad \tau_k = 1$$

$$\text{otherwise } \tau_k = \min (\|g_k\|^3 / (\Delta_k g_k^T B_k g_k), 1)$$



## Limitations:

- Though Cauchy point is cheap to implement, it performs poorly in some cases.
- Various kinds of improvements are based on including the curvature information from  $B_k$ .

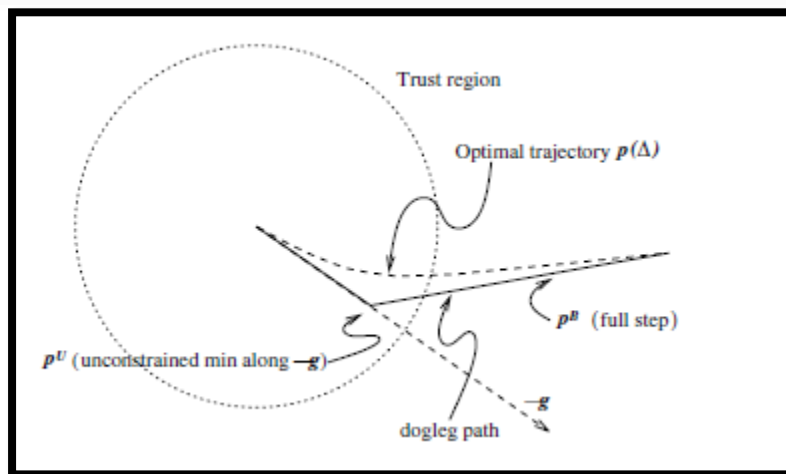
# Dogleg method

If  $B_k$  is positive definite (we can use quasi-Newton Hessian approximation & updating to guarantee), then a V-shaped trajectory can be determined by

if  $0 \leq \tau \leq 1$ ,  $p(\tau) = \tau p^U$

if  $1 \leq \tau \leq 2$ ,  $p(\tau) = \tau p^U + (\tau - 1)(p^B - p^U)$

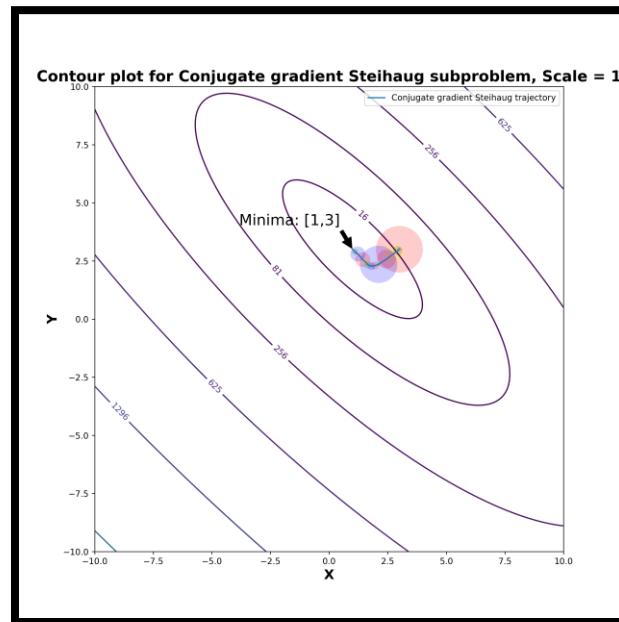
where  $p^U = -\frac{g^T g}{g^T B g} g$  is the steepest descent direction and  $p^B$  is the optimal solution of the quadratic model  $m_k(p)$ . Therefore, a further improvement could be achieved compared to using only Cauchy point calculation method in one iteration. (Note that hessian or approximate hessian will be evaluated in dogleg method)



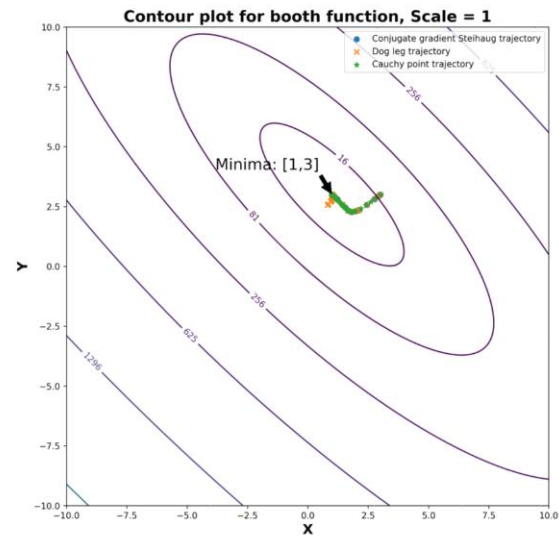
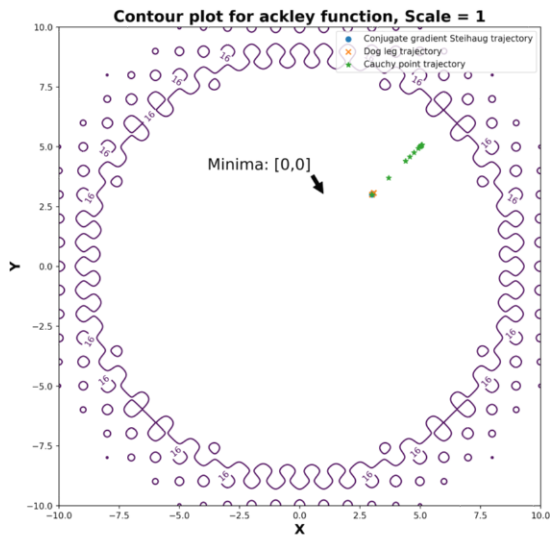
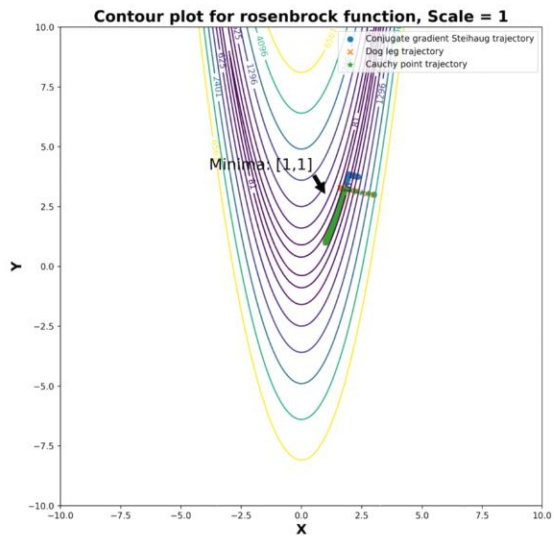
# Conjugate Gradient Steihaug method

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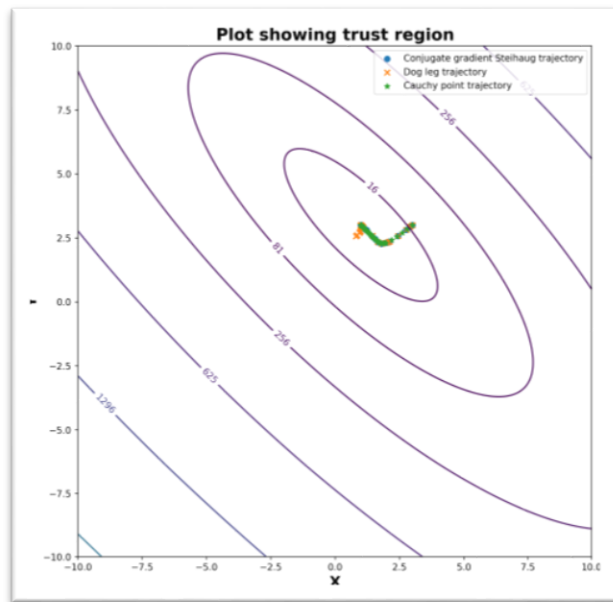
Given tolerance  $\epsilon_k > 0$ ;
Set  $z_0 = 0, r_0 = \nabla f_k, d_0 = -r_0 = -\nabla f_k$ 
if  $\|r_0\| < \epsilon_k$ 
return  $p_k = z_0 = 0$ ;
for  $j = 0, 1, 2, \dots$ 
if  $d_j^T B_k d_j \leq 0$ 
Find  $\tau$  such that  $p_k = z_j + \tau d_j$  minimizes  $m_k(p_k)$ 
and satisfies  $\|p_k\| = \Delta_k$ ;
return  $p_k$ ;
Set  $\alpha_j = r_j^T r_j / d_j^T B_k d_j$ ;
Set  $z_{j+1} = z_j + \alpha_j d_j$ ;
if  $\|z_{j+1}\| \geq \Delta_k$ 
Find  $\tau \geq 0$  such that  $p_k = z_j + \tau d_j$  satisfies  $\|p_k\| = \Delta_k$ ;
return  $p_k$ ;
Set  $r_{j+1} = r_j + \alpha_j B_k d_j$ ;
if  $\|r_{j+1}\| < \epsilon_k$ 
return  $p_k = z_{j+1}$ ;
Set  $\beta_{j+1} = \frac{r_{j+1}^T r_{j+1}}{r_j^T r_j}$ ;
Set  $d_{j+1} = -r_{j+1} + \beta_{j+1} d_j$ 
end
    
```



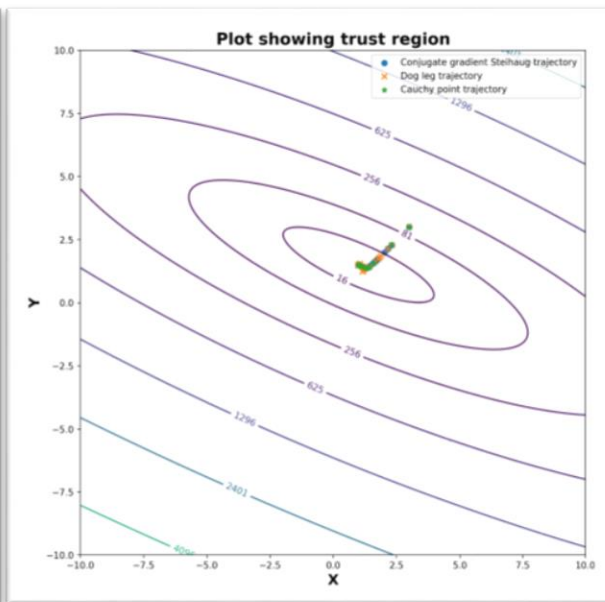
# Result: Contour Plot



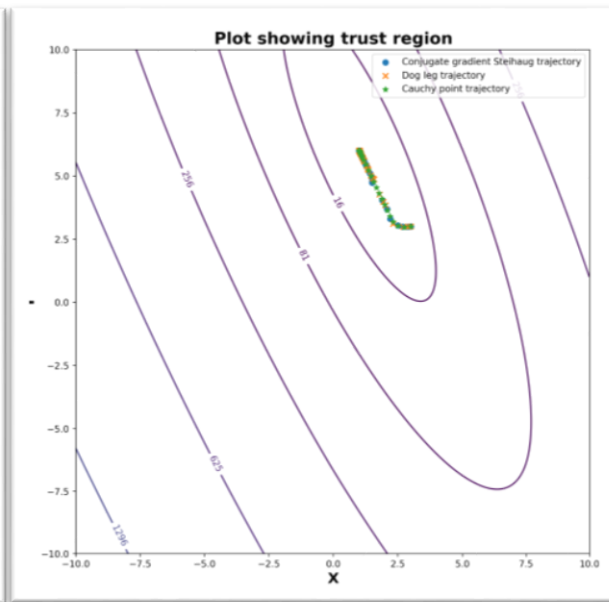
# Result: Scaling (Booth function)



Scale=1.0



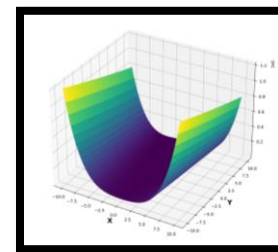
Scale=2.0



Scale=0.5



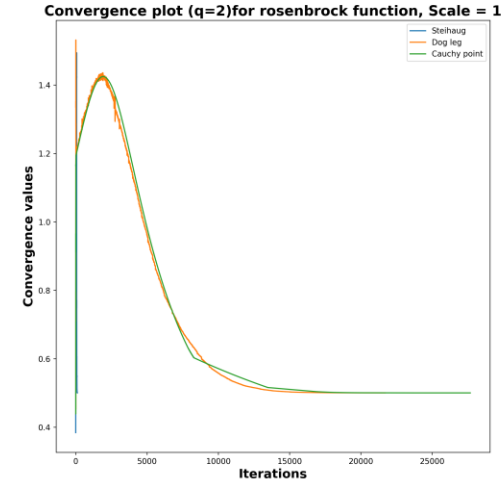
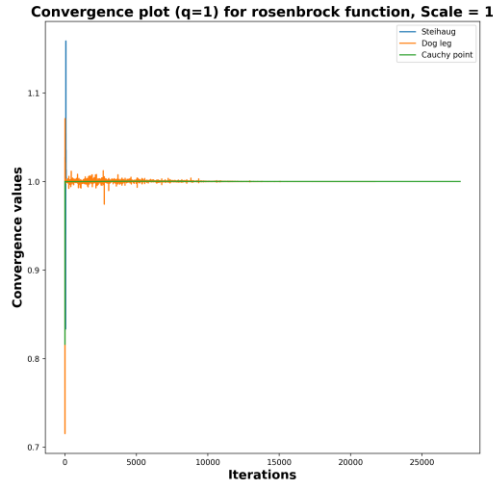
# Result: Analysis (Rosenbrock)



Initial parameters:  $x_0 = (3, 3)$ ;  $\Delta_{max} = 5$ ;  $\Delta_0 = 1$

	Cauchy point	Dogleg method	CG Steihaug
Number of iterations	27678	21074	118
Execution time	2.915 sec	9.714 sec	0.035 sec
Optimal point $x_k$	(1.00000433, 1.00000868)	(1.0001110, 1.00022252)	(1.00007155, 1.00014345)
Scale=0.5 (iterations)	30000	30000	21
Scale=1.0 (iterations)	27678	21074	118
Scale=2.0 (iterations)	25340	22686	3301

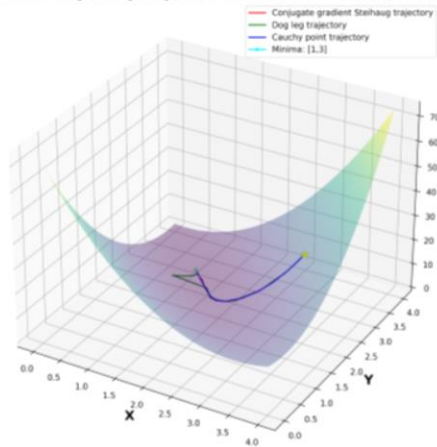
# Result: Convergence (Rosenbrock)



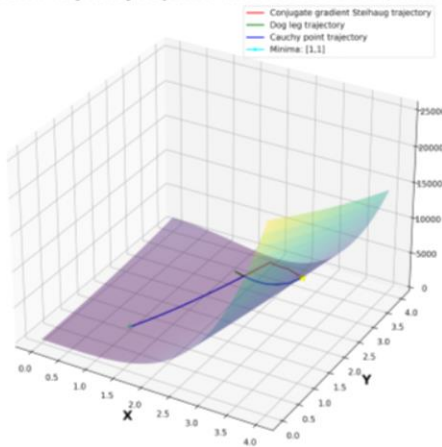
Subproblem solution methods	Rosenbrock 'q' value	Booth 'q' value	Ackley 'q' value
Cauchy point	1.000127	0.991538	0.99999
Dog Leg	1.00008	1.008220	1.00000
Steihaug	0.986129	1.011392	0.01662

# Result: Trajectory 3D plot

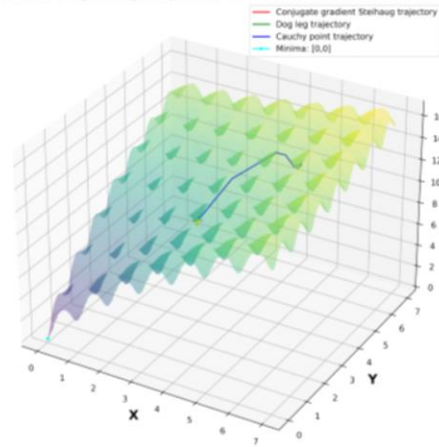
3D Plot with trajectory of points towards the minima, Scale = 1



3D Plot with trajectory of points towards the minima, Scale = 1



3D Plot with trajectory of points towards the minima, Scale = 1



# Conclusion

- Variants to solve trust region subproblem was investigated.
- Plots describing the optimization process were made.
- Effect of scaling on the objective functions, run time of the subproblem variants were analysed.
- The convergence of subproblem solutions variants was investigated.

# References

1. Numerical optimization lecture manuscript by Prof. Moritz Diehl.
2. [Jorge Nocedal and Stephen J. Wright, Numerical Optimization, Springer, 2006](#)
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4. Reference trust region code and plot functions: [https://github.com/anweshpanda/Trust\\_Region/blob/main/Trust%20Region.ipynb](https://github.com/anweshpanda/Trust_Region/blob/main/Trust%20Region.ipynb)
5. Steihaug method theory: <https://vdocuments.mx/the-dogleg-and-steihaug-methods-science-and-technology-the-dogleg-and.html?page=22>
6. Trust region theory and pseudo code: [https://web.archive.org/web/20220922103636/https://optimization.mccormick.northwestern.edu/index.php/Trust-region\\_methods](https://web.archive.org/web/20220922103636/https://optimization.mccormick.northwestern.edu/index.php/Trust-region_methods)

**Thank you for your attention!**