

	Gradient	Subgrad	Prox grad	Newton	quasi-Newton
Criterion	smooth f	any f	smooth + simple, $f = g + h$	doubly smooth f	doubly smooth f
Constraints	projection onto con- straint set	projection onto con- straint set	constrained prox opera- tor	equality con- straints	unconstrained
Opti pa- rameters	fixed step size ($t \leq 1/L$) or line search	diminishing step sizes	fixed step size ($t \leq 1/L$) or line search	pure step size ($t = 1$) or line search	line search
Iteration cost	cheap (compute gradient)	cheap (compute subgradi- ent)	moderately cheap (evaluate prox)	moderate to expensive (compute Hes- sian and solve linear system)	moderately cheap (com- pute gradients, inner products; no matrix inversion)
Rate	$O(\frac{1}{\epsilon})$ [$O(\frac{1}{\sqrt{\epsilon}})$ with ac- celeration, $O(\log(\frac{1}{\epsilon}))$ with strong convexity]	$O(\frac{1}{\epsilon^2})$	$O(\frac{1}{\epsilon})$ [$O(\frac{1}{\sqrt{\epsilon}})$ with ac- celeration, $O(\log(\frac{1}{\epsilon}))$ with strong convexity]	$O(\log \log(\frac{1}{\epsilon}))$ (local quadratic convergence rate)	superlinear rate, n -step quadratic rate (n steps are as effective as one Newton step)

	Barrier method	Primal-dual IPM	ADMM	Coord descent
Criterion	doubly smooth f	doubly smooth f	block separable, $f(x, z) = g(x) + h(z)$	smooth + coordinate-wise separable
Constraints	doubly smooth h_i (inequality constraints)	doubly smooth h_i (inequality constraints)	eq constraints (always) & ineq constraints (sometimes)	coordinate- wise separable constraints
Opti pa- rameters	inner loop: fixed step size or use line search, outer loop: diverging barrier parameter	line search for step size & diverging barrier parameter	fixed aug- mented La- grange paramete- ter (theory), or varied by itera- tion (practice)	none!
Iteration cost	expensive to very expensive (one iteration solves one smoothed problem, by Newton)	moderate to expensive (one iteration per- forms one Newton step)	cheap to ex- pensive (one iteration solves two subprob- lems, makes a dual step)	cheap to expen- sive (one iter- ation performs a full cycle of coordinate min- imizations)
Rate	$O(\log(\frac{1}{\epsilon}))$ (both in terms of iterations and Newton steps)	$O(\log(\frac{1}{\epsilon}))$	not known in general, known in special cases; practically tends to behave like first-order methods	not known in general, known in special cases; practically tends to behave faster than first-order methods