

Barrier functions

In essence, barrier methods also use the notion of using proxies for the constraints in the objective function, so that an unconstrained optimisation problem can be solved instead. However, the concept of barrier functions is different than penalty functions in that they are defined to prevent the solution search method from leaving the feasible region, which is why some of these methods are also called interior point methods.

Consider the primal problem P being defined as $\text{flalign}^*(P) : \min_x f(x)$

We define the barrier problem BP as $\text{flalign}^*(BP) : \min_{\mu} \theta(\mu)$

where $\theta(\mu) = \inf_x f(x) + \mu B(x) : g(x) < 0, x \in X$ and $B(x)$ is a barrier function. The barrier function is such that its value approaches $+\infty$ as the boundary of the region $x : g(x) \leq 0$ is approached from its interior. Notice that, in practice, it means that the constraint $g(x) < 0$ can be dropped, as they are automatically enforced by the barrier function.

The barrier function $B : \mathbb{R}^n \rightarrow \mathbb{R}$ is such that equation $B(x) = \sum_{i=1}^m \phi(g_i(x))$, where $\phi(y) \geq 0$, if $y < 0$;

Perhaps the most important barrier function is the Frisch's log barrier function, used in the highly successful primal-dual interior point methods. We will describe its use later. The log barrier is defined as

$$B(x) = - \sum_{i=1}^m \ln(-g_i(x)).$$

Figure fig:different_m illustrates the behaviour of the barrier function. Ideally, the barrier function $B(x)$ has the role of forcing $x : g(x) < 0$ but assume infinite value if a solution is at the boundary $g(x) = 0$ or outside the feasible region. This is illustrated in the dashed line in Figure fig:different_mu. The barrier functions for different values of barrier term μ illustrate how the log barrier mimics this behaviour, becoming more and more pronounced as μ decreases.

figure [width=]part₂/chapter₁₀/figures/different_mu.pdf The barrier function for different values of μ
fig:different_mu