In binary classification problems it is common for the data sets to be very imbalanced: one class is very rare compared to the other. In this work we consider the infinitely imbalanced case where the rare class has fixed finite sample size $n$, while the common class has sample size $N \to \infty$. For logistic regression, the infinitely imbalanced case often has a useful solution. The logistic regression intercept typically diverges to $-\infty$ as expected. But under mild conditions, the rest of the coefficient vector approaches a non trivial, interpretable and useful limit. Perhaps surprisingly, the limiting parameter vector depends on the $n$ points from the rare class only through their sample mean.