# IEEE TEX.jl Example

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## I. Loss Function

In mathematical optimization, statistics, decision theory and machine learning, a *loss function* or *cost function* is a function that maps an event or values of one or more variables onto a real number intuitively representing some "cost" associated with the event.[1] An optimization problem seeks to minimize a loss function. An objective function is either a loss function or its negative (sometimes called a *reward function* or a *utility function*), in which case it is to be maximized.

$$J(\theta) = \frac{1}{m}\sum_{i=1}^{m}\Big[-y_i \log(h_\theta(x_i)) - (1-y_i)\log(1-h_\theta(x_i))\Big] \tag{1}$$

```
function loss_function(theta, X, y)
    m = length(y) # number of training examples
    grad = zeros(size(theta))
    h = sigmoid(X * theta)
    J = 1/m * sum((-y'*log(h))-(1 .- y)'*log(1 .- h))
    grad = 1/m*(X'*(h-y))
    return (J, grad)
end
```

[1] https://en.wikipedia.org/wiki/Loss_function