

LAB #8: FINDING DESCENT TO MINIMIZE

Recall, we want to minimize

$$F(u) = \int_{\Omega} |\nabla u|^p dx + \lambda \int_{\Omega} |f(x) - u(x)|^q dx.$$

Let's first put together some thoughts about minimizing:

- (1) Given a function, $f : \mathbb{R}^n \rightarrow \mathbb{R}$, at a point $a \in \mathbb{R}^n$, in which direction should you travel to find greatest descent?
- (2) Suppose you follow this descent direction, why would you need to go a short distance and then reevaluate the descent direction? What could happen if you took a large step along this direction?
- (3) Draw an example of a function where the scenario you described above would happen.
- (4) In optimization terminology, looking along a line parallel to a descent direction to a "better" point is called a "line search." As we said in the last problem, if we take too big of a step along this line, we may have a problem. Because we know the direction in which we step is a direction of descent, what can we do to be sure we reach a better point?
- (5) Following the steps below, we get a sequence (x_n) of better points.
 - (a) At a current point, x_k find the direction of descent d_k (described above)
 - (b) Use a line search to find a better point, one where f has a lower value.
 - (c) Name the better point x^* .
 - (d) Repeat for $x_k = x^*$.

How can this plan go wrong?

- (6) List two ideas to combat the problems that might happen with the above algorithm. Describe how these ideas combat the issues in 5.
- (7) When looking for a minimizer of a function, what must be true about the gradient at that point?
- (8) We found that the minimizer, \bar{u} of F must satisfy the equation

$$p\nabla \cdot (|\nabla u|^{p-2}\nabla u) - \lambda q|f - u|^{q-2}(f - u) = 0.$$

How did we go about getting this equation?

- (9) Using the last two questions, what can we use as the "gradient" of $F(u)$?
- (10) Use your answers above to determine a direction of fastest descent at a point u .