# Math Camp

Module #5: Functions on  $\mathbb{R}^n$ 

#### $\mathbb{R}^n$

A *Hilbert space* is a complete, normed linear space (i.e. Banach space) in which the norm is induced by an inner product.

 $\mathbb{R}^n$  is a Hilbert space, with inner product given by the usual dot product:

$$||x||_2 = \sqrt{\langle x, x \rangle} = \sqrt{\sum_{i=1}^n x_i^2}$$

## Continuous functions

- Same as before: continuous functions preserve convergent sequences.
- On R the main result was the IVT
- Recall that  $K \subset \mathbb{R}^n$  is *convex* if  $x, y \in K \implies \alpha x + (1 \alpha)y \in K$

## Theorem (Brouwer)

Let  $K \subset \mathbb{R}^n$  be compact and convex, and let  $f: K \to K$  be continuous. Then there is a point  $x \in K$  such that f(x) = x.

Remark: this is an existence theorem

## Continuous functions

 $f: \mathbb{R}^n \to \mathbb{R}^n$  is a *contraction* if there is a  $\beta \in (0,1)$  such that for any  $x,y \in \mathbb{R}^n$  we have that

$$||f(x) - f(y)|| \le \beta ||x - y||.$$

# Theorem (Contraction mapping theorem)

Let  $X \subset \mathbb{R}^n$  be closed. Suppose  $f: X \to X$  is a contraction. Then there exists a unique  $x^* \in X$  such that  $f(x^*) = x^*$ . Furthermore, if  $x_0$  is any point in X then

$$f^{n}(x_{0}) \equiv f(f^{n-1}(x_{0})) \to x^{*}.$$

Remark: this theorem is constructive!

## Continuous functions

## Recall:

# Theorem (Attainment of extrema)

Let  $K \subset \mathbb{R}^n$  be compact and  $f: K \to \mathbb{R}$  be continuous. Then there exists  $x \in K$  such that

$$f(x^*) = \sup_{x \in K} f(x) < \infty.$$

#### Partial derivatives

Let  $f : \mathbb{R}^n \to \mathbb{R}$ . Then the partial derivative of f with respect to  $x_i$  at  $x \in \mathbb{R}^n$ , when it it exists, is computed as

$$f_i(x) = \lim_{\Delta x_i \to 0} \frac{f(x_1, \dots, x_{i-1}, x_i + \Delta x_i, x_{i+1}, \dots, x_n) - f(x)}{\Delta x_i}.$$

Remark:  $f_{ij}$  is the partial derivative of  $f_i$  with respect to  $x_j$ . If  $f_{ij}$  and  $f_{ji}$  are continuous then they are equal.

## Total differentials

if  $f: \mathbb{R}^n \to \mathbb{R}$  then the *total differential* of f is

$$df = \sum_{i=1}^{n} f_i \, dx_i = f_i \, dx_i$$

Example: Use the Keynesian cross to determine the impact on output of a rise in taxes.

#### **Derivatives**

- Let  $f: \mathbb{R}^n \to \mathbb{R}$  and  $dx = (dx_1, \dots, dx_n)^T$
- Let Df be the *row* vector  $(f_1, \ldots, f_n)$  (linear functional!)
- $df = \sum_{i=1}^{n} f_i dx_i = Df \cdot dx$
- More generally, if  $f: \mathbb{R}^n \to \mathbb{R}^m$  then  $Df(x): \mathbb{R}^m \to \mathbb{R}^m$  is the unique linear map satisfying

$$\lim_{\|\Delta x\| \to 0} \frac{\|f(x+\Delta x) - f(x) - Df(x)(\Delta x)\|}{\|\Delta x\|} = 0$$

## **Derivatives**

If 
$$f: \mathbb{R}^n \to \mathbb{R}^m$$
 write  $f = \begin{pmatrix} f^1 \\ \vdots \\ f^m \end{pmatrix}$ 

# Theorem (The Jacobian)

If  $f: \mathbb{R}^n \to \mathbb{R}^m$  is differentiable and  $f_i^i$  is continuous then

$$Df(x) = \begin{pmatrix} f_1^1 & \cdots & f_n^1 \\ \vdots & \ddots & \vdots \\ f_1^m & \cdots & f_n^m \end{pmatrix}$$

## **Derivatives**

Notice that  $f: \mathbb{R}^n \to \mathbb{R}$  implies

$$df = \begin{pmatrix} df^1 \\ \vdots \\ df^m \end{pmatrix} = \begin{pmatrix} Df^1 dx \\ \vdots \\ Df^m dx \end{pmatrix} = \begin{pmatrix} Df^1 \\ \vdots \\ Df^m \end{pmatrix} dx = Df \cdot dx$$

# Taylor's theorem: univariate

If  $f: \mathbb{R} \to \mathbb{R}$  is  $C^{N+1}$  then

$$f(x) = f(x^*) + \sum_{n=1}^{N} \frac{f^{(n)}(x^*)}{n!} (x - x^*)^n + \mathcal{O}(|x - x^*|^{N+1}).$$

## Taylor's theorem: multivariate

Let  $f: \mathbb{R}^n \to \mathbb{R}$ .

- $Df(x) \in (\mathbb{R}^n)^* \cong \mathbb{R}^n$ .
- Thus  $Df: \mathbb{R}^n \to \mathbb{R}^n$ .
- So  $D^2 f(x) : \mathbb{R}^n \to \mathbb{R}^n$  and  $D^2 f : \mathbb{R}^n \to \mathbb{R}^{n \times n}$ .
- The *Hessian* is  $D^2f$ , which is the matrix of second partials.
- Taylor's theorem

$$f(x) = f(x^*) + Df(x^*)(x - x^*) + \frac{1}{2}(x - x^*)^T D^2 f(x^*)(x - x^*) + \mathcal{O}\left(\|x - x^*\|^3\right)$$

# Comparative statics

Comparative statics: let  $f: \mathbb{R}^m \oplus \mathbb{R}^n \to \mathbb{R}^n$  and consider the equation

$$f(x,y) = 0 \in \mathbb{R}^n$$

If x moves by dx, how must y move (dy) so that

$$f(x+dx, y+dy) = 0?$$

## Implicit function theorem

Let  $f: \mathbb{R}^m \oplus \mathbb{R}^n \to \mathbb{R}^n$  and define the matrix partials

$$D_x f(x,y) = \begin{pmatrix} f_{x_1}^1(x,y) & \cdots & f_{x_m}^1(x,y) \\ \vdots & \ddots & \vdots \\ f_{x_1}^n(x,y) & \cdots & f_{x_m}^n(x,y) \end{pmatrix}$$

$$D_{y}f(x,y) = \begin{pmatrix} f_{y_{1}}^{1}(x,y) & \cdots & f_{y_{n}}^{1}(x,y) \\ \vdots & \ddots & \vdots \\ f_{y_{1}}^{n}(x,y) & \cdots & f_{y_{n}}^{n}(x,y) \end{pmatrix}$$

## Theorem (Implicit function theorem)

Suppose  $f: \mathbb{R}^m \oplus \mathbb{R}^n \to \mathbb{R}^n$  has continuous first partials. If  $f(x^*, y^*) = 0$  and  $\det D_y f(x^*, y^*) \neq 0$  then there exists open set  $U \subset \mathbb{R}^m$ , with  $x^* \in U$ , and a continuously differentiable function  $g: U \to \mathbb{R}^n$  so that

- $y^* = g(x^*)$
- $x \in U \implies f(x, g(x)) = 0$
- $Dg(x^*) = -D_y f(x^*, y^*)^{-1} \circ D_x f(x^*, y^*)$