

## CAL III: FUNCTIONS & GRADIENT

### Functions of Several Variables:

- Just as a typical one-dimensional function maps an input ( $x$ , horizontal axis) to an output ( $y$ , vertical axis), so a two-dimensional function maps inputs ( $x$  and  $y$ ) to an output ( $z$ , height). All rules still apply to higher-order functions and many properties scale.

### Directional Derivatives and Gradient:

- Directional derivatives describe how rapidly a function is changing in a certain direction as defined by the unit vector of  $v$ . A typical partial derivative is a specific directional derivative, with the unit vector in the  $i$  or  $j$  direction. If you are given a vector  $\mathbf{u}$  that is not a unit vector, then the unit vector  $\mathbf{v}$  would be  $\mathbf{u}/\|\mathbf{u}\|$ .
- Gradients define the rate of change in the direction that water would flow along the surface; that is, the direction of greatest increase/decrease at a specific point. A vector orthogonal to the gradient experiences no change. This is analogous to a contour map.

The gradient of  $f$  is noted by the Greek letter "del":

$$\nabla f = \frac{\partial f}{\partial x}i + \frac{\partial f}{\partial y}j$$

The directional derivative at point  $P_0$  in the

direction of  $u = \frac{v}{|v|}$ :

$$\left(\frac{\partial f}{\partial s}\right)\Big|_{u, P_0} = (\nabla f)\Big|_{P_0} \cdot u$$

### Extreme Values:

- First, determine the location  $(a, b)$  such that  $f_x = f_y = 0$ . *Note: this often requires factoring.*
- Next, apply the second derivative test to determine which feature is located at  $(a, b)$ .
- If the discriminant of  $f$  is  $< 0$  (negative),  $f$  has a **saddle point** at  $(a, b)$ .
- If the discriminant of  $f$  is  $= 0$ , the test is **inconclusive**.
- If the discriminant of  $f$  is  $> 0$  (positive), evaluate the following rules:
  - If  $f_{xx} < 0$  (negative) then there is a **local maximum** at  $(a, b)$ .
  - If  $f_{xx} > 0$  (positive) then there is a **local minimum** at  $(a, b)$ .
- The discriminant is defined as  $f_{xx} \cdot f_{yy} - f_{xy}^2$ , which can be remembered using  $\left| \begin{matrix} f_{xx} & f_{xy} \\ f_{xy} & f_{yy} \end{matrix} \right|$