

# ME451

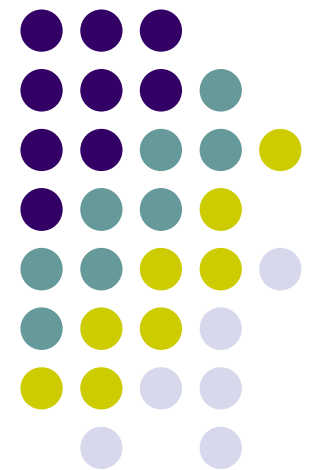
## Kinematics and Dynamics of Machine Systems

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Review of Elements of Calculus – 2.5

Vel and Acc of a Point fixed in a Ref Frame – 2.6

September 14, 2010

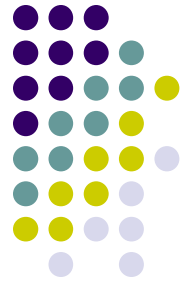


# Before we get started...



- **NOTE:**
  - Next Tu and Th you'll go through two ADAMS tutorials
  - TA to email to lab where you'll meet for these two tutorials
- HW Assigned (due next Tu):
  - 2.4.4, 2.5.1, 2.5.2, 2.5.3, 2.5.7
  - ADAMS component to be emailed to you by TA
- Last time:
  - Discussed about Recursive and Absolute Generalized Coordinates
    - Each has its own advantages/disadvantages
  - Brief linear algebra review
- Today:
  - Time derivatives of vectors and matrices
  - Computing the partial derivate of functions
  - Discuss chain rule for taking time derivatives
  - Computing the velocity and acceleration of a point attached to a moving rigid body

# Derivatives of Functions



- GOAL: Understand how to
  - Take time derivatives of vectors and matrices
  - Take partial derivatives of functions with respect to its arguments
    - We will use a matrix-vector notation for computing these partial derivs.
    - Taking partial derivatives might be challenging in the beginning
    - It will be used a lot in this class

# Taking time derivatives of a time dependent vector



- FRAMEWORK:

- Vector  $\mathbf{r}$  is represented as a function of time, and it has two components:  $x(t)$  and  $y(t)$ :

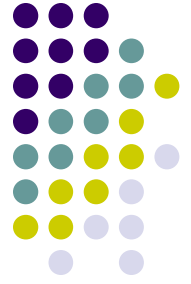
$$\mathbf{r}(t) = \begin{bmatrix} x(t) \\ y(t) \end{bmatrix}$$

- Its components change, but the vector is represented in a **fixed** reference frame

- THEN:

$$\dot{\mathbf{r}}(t) = \begin{bmatrix} \dot{x}(t) \\ \dot{y}(t) \end{bmatrix}, \quad \ddot{\mathbf{r}}(t) = \begin{bmatrix} \ddot{x}(t) \\ \ddot{y}(t) \end{bmatrix}, \quad \text{etc.}$$

# Time Derivatives, Vector Related Operations



- Assume that  $\alpha \in \mathbb{R}$ ,  $\mathbf{a} \in \mathbb{R}^n$ ,  $\mathbf{b} \in \mathbb{R}^n$  depend on time. Then it can be proved that the following hold:

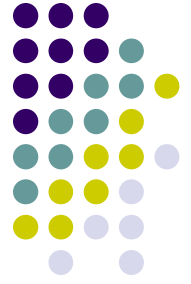
$$\frac{d}{dt}(\alpha \mathbf{a}) = \frac{d\alpha}{dt} \mathbf{a} + \alpha \frac{d\mathbf{a}}{dt} = \dot{\alpha} \mathbf{a} + \alpha \dot{\mathbf{a}}$$

$$\frac{d}{dt}(\mathbf{a} + \mathbf{b}) = \frac{d\mathbf{a}}{dt} + \frac{d\mathbf{b}}{dt} = \dot{\mathbf{a}} + \dot{\mathbf{b}}$$

$$\frac{d}{dt}(\mathbf{a}^T \mathbf{b}) = \frac{d\mathbf{a}^T}{dt} \mathbf{b} + \mathbf{a}^T \frac{d\mathbf{b}}{dt} = \dot{\mathbf{a}}^T \mathbf{b} + \mathbf{a}^T \dot{\mathbf{b}}$$

$$\mathbf{a}^T \mathbf{a} = \text{const} \quad \Rightarrow \quad \mathbf{a}^T \dot{\mathbf{a}} = 0$$

# Taking time derivatives of MATRICES

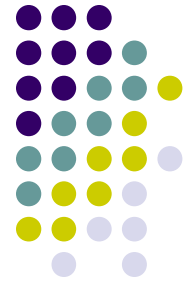


- By definition, the time derivative of a matrix is obtained by taking the time derivative of each entry in the matrix
- A simple extension of what we've seen for vector derivatives
- Assume that  $\alpha \in \mathbb{R}$ ,  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\mathbf{B} \in \mathbb{R}^{m \times n}$ , and  $\mathbf{C} \in \mathbb{R}^{n \times p}$  depend on time. Then it can be proved that the following hold:

$$\frac{d}{dt}(\alpha \mathbf{A}) = \frac{d\alpha}{dt} \mathbf{A} + \alpha \frac{d\mathbf{A}}{dt} = \dot{\alpha} \mathbf{A} + \alpha \dot{\mathbf{A}}$$

$$\frac{d}{dt}(\mathbf{A} + \mathbf{B}) = \frac{d\mathbf{A}}{dt} + \frac{d\mathbf{B}}{dt} = \dot{\mathbf{A}} + \dot{\mathbf{B}}$$

$$\frac{d}{dt}(\mathbf{A}\mathbf{C}) = \frac{d\mathbf{A}}{dt} \mathbf{C} + \mathbf{A} \frac{d\mathbf{C}}{dt} = \dot{\mathbf{A}}\mathbf{C} + \mathbf{A}\dot{\mathbf{C}}$$



End Time Derivatives  
...  
Discuss Partial Derivatives

# What's the story behind the concept of partial derivative?



- What's the meaning of a partial derivative?
  - It captures the “sensitivity” of a function quantity wrt a variable that the function depends upon
  - Shows how much the function changes when the variable changes a bit
- Simplest case of partial derivative: you have one function that depends on one variable:

$$f(x) = \ln x \quad , \quad g(z) = \sin(4z + \pi) \quad , \quad \textit{etc.}$$

- Then,

$$\frac{\partial f}{\partial x} = \frac{1}{x} \quad , \quad \frac{\partial g}{\partial z} = 4 \cos(4z + \pi) \quad , \quad \textit{etc.}$$



# Partial Derivative, Two Variables



- Suppose you have one function but it depends on **two** variables, say  $x$  and  $y$ :

$$f(x, y) = \sin(x^2 + 3y^2)$$

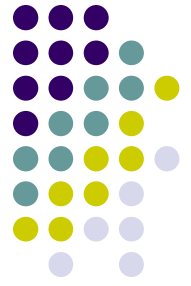
- To simplify the notation, an array  $\mathbf{q}$  is introduced:

$$\mathbf{q} = \begin{bmatrix} x \\ y \end{bmatrix} \in \mathbb{R}^2$$

- With this, the partial derivative of  $f$  wrt  $\mathbf{q}$  is defined as

$$\left[ \frac{\partial f}{\partial x} \quad \frac{\partial f}{\partial y} \right] \equiv f_{x,y} \equiv \frac{\partial f}{\partial \mathbf{q}} \equiv f_{\mathbf{q}} = [2x \cos(x^2 + 3y^2) \quad 6y \cos(x^2 + 3y^2)]$$

## ...and here is as good as it gets (vector function)



- You have a group of “m” functions that are gathered together in an array, and they depend on a collection of “n” variables:

$f_1, f_2, \dots, f_m$  depend on  $x_1, x_2, \dots, x_n$

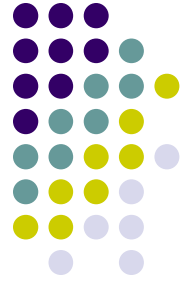
- The array that collects all “m” functions is called **F**:

$$\mathbf{F}(x_1, x_2, \dots, x_n) = \begin{bmatrix} f_1(x_1, x_2, \dots, x_n) \\ f_2(x_1, x_2, \dots, x_n) \\ \dots \\ f_m(x_1, x_2, \dots, x_n) \end{bmatrix} \in \mathbb{R}^m$$

- The array that collects all “n” variables is called **q**:

$$\mathbf{q} = \begin{bmatrix} x_1 \\ x_2 \\ \dots \\ x_n \end{bmatrix} \in \mathbb{R}^n$$

# Most general partial derivative (Vector Function, Cntd.)



- Then, in the most general case, we have  $\mathbf{F}(\mathbf{q})$ , and

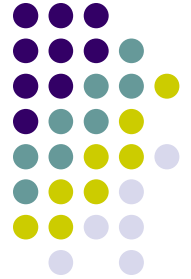
$$\frac{\partial \mathbf{F}}{\partial \mathbf{q}} \equiv \mathbf{F}_{\mathbf{q}} \equiv \begin{bmatrix} \frac{\partial f_1}{\partial \mathbf{q}} \\ \frac{\partial f_2}{\partial \mathbf{q}} \\ \dots \\ \frac{\partial f_m}{\partial \mathbf{q}} \end{bmatrix} = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \frac{\partial f_2}{\partial x_n} \\ \dots & \dots & \dots & \dots \\ \frac{\partial f_m}{\partial x_1} & \frac{\partial f_m}{\partial x_2} & \dots & \frac{\partial f_m}{\partial x_n} \end{bmatrix} \in \mathbb{R}^{m \times n}$$

This is an  $m \times n$  **matrix!**

- Example 2.5.2:

$$\mathbf{q} = \begin{bmatrix} \theta_1 \\ \theta_2 \end{bmatrix} \quad \mathbf{r}^P = \begin{bmatrix} \cos \theta_1 + l \cos(\theta_1 + \theta_2) \\ \sin \theta_1 + l \sin(\theta_1 + \theta_2) \end{bmatrix} \quad \mathbf{r}_{\mathbf{q}}^P = ?$$

# A Word on Notation: Left and Right mean the same thing



- Let  $x$ ,  $y$ , and  $\phi$  be three generalized coordinates
- Define the function  $\mathbf{r}$  of  $x$ ,  $y$ , and  $\phi$  as

$$\mathbf{r}(x, y, \phi) = \begin{bmatrix} x + 2l\cos\phi \\ y - 2l\sin\phi \end{bmatrix}$$

- Verbose notation

$$\mathbf{r}_{x,y,\phi} = \begin{bmatrix} \frac{\partial \mathbf{r}}{\partial x} & \frac{\partial \mathbf{r}}{\partial y} & \frac{\partial \mathbf{r}}{\partial \phi} \end{bmatrix}$$

- Let  $x$ ,  $y$ , and  $\phi$  be three generalized coordinates, and define the array  $\mathbf{q}$

$$\mathbf{q} = \begin{bmatrix} x \\ y \\ \phi \end{bmatrix}$$

- Define the function  $\mathbf{r}$  of  $\mathbf{q}$ :

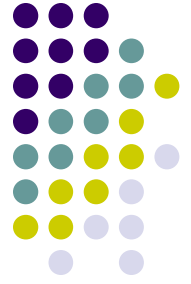
$$\mathbf{r}(\mathbf{q}) = \begin{bmatrix} x + 2l\cos\phi \\ y - 2l\sin\phi \end{bmatrix}$$

- Terse notation

$$\mathbf{r}_{\mathbf{q}} = \frac{\partial \mathbf{r}}{\partial \mathbf{q}}$$



# Example



$$\mathbf{q} = \begin{bmatrix} x \\ y \\ \phi \end{bmatrix}$$

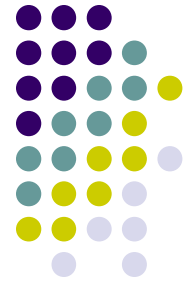
$$\mathbf{r}(\mathbf{q}) = \begin{bmatrix} x + 2l \cos \phi \\ y - 2l \sin \phi \end{bmatrix}$$

$$\mathbf{r}_{\mathbf{q}} = \frac{\partial \mathbf{r}}{\partial \mathbf{q}} = ?$$

$$\mathbf{r}_{\mathbf{q}} = \frac{\partial \mathbf{r}}{\partial \mathbf{q}} = \left[ \frac{\partial \mathbf{r}}{\partial q_1} \quad \frac{\partial \mathbf{r}}{\partial q_2} \quad \frac{\partial \mathbf{r}}{\partial q_3} \right] = \left[ \frac{\partial \mathbf{r}}{\partial x} \quad \frac{\partial \mathbf{r}}{\partial y} \quad \frac{\partial \mathbf{r}}{\partial \phi} \right]$$

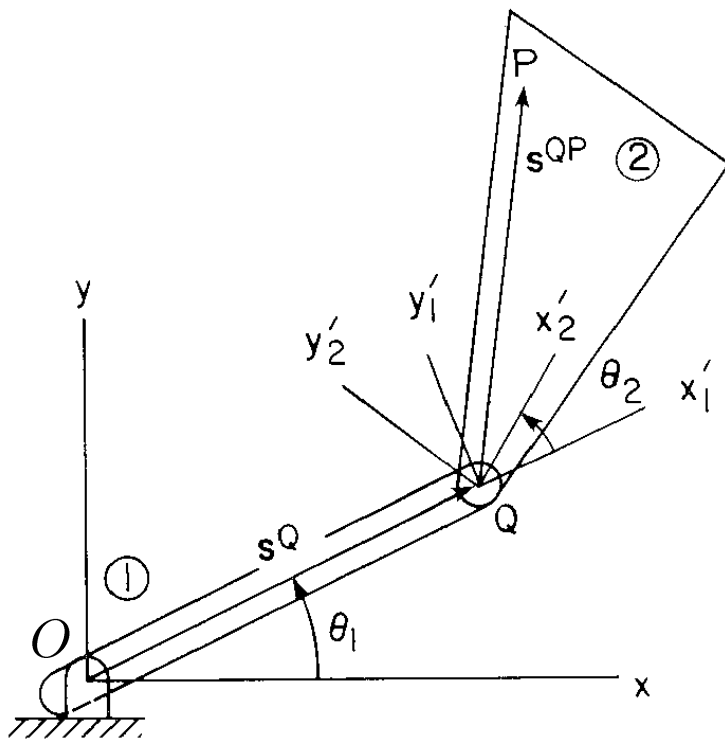


$$\mathbf{r}_{\mathbf{q}} = \begin{bmatrix} 1 & 0 & -2l \sin \phi \\ 0 & 1 & -2l \cos \phi \end{bmatrix}$$



## Another Example (builds on Example 2.4.1)

- Let  $\mathbf{q} = \begin{bmatrix} \theta_1 \\ \theta_2 \end{bmatrix}$
- Find the partial derivative of the position of P with respect to the array of generalized coordinates  $\mathbf{q}$



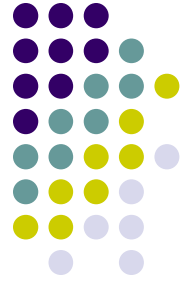
$$s'^P = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$$

$$\|\overrightarrow{OQ}\| = 5$$

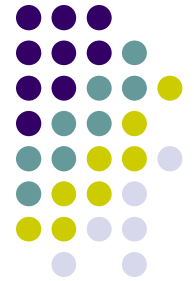
$$\mathbf{r}_{\mathbf{q}}^P = \frac{\partial \mathbf{r}^P}{\partial \mathbf{q}} = \left[ \frac{\partial \mathbf{r}^P}{\partial \theta_1} \quad \frac{\partial \mathbf{r}^P}{\partial \theta_2} \right] = ?$$

Figure 2.4.5 Two-body positioning mechanism.

# Partial Derivatives: Good to Remember...



- In the most general case, you start with “m” functions in “n” variables, and end with an  $(m \times n)$  matrix of partial derivatives.
  - You start with a column vector of functions and then end up with a matrix
- Taking a partial derivative leads to a *higher dimension* quantity
  - Scalar Function – leads to row vector
  - Vector Function – leads to matrix
  - I call this the “accordion rule”
- In this class, taking partial derivatives can lead to one of the following:
  - A row vector
  - A full blown matrix
  - If you see something else chances are you made a mistake...



Done with Partial Derivatives

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Moving on to Chain Rule of Differentiation



# Scenario 1: Scalar Function



- $f$  is a function of “ $n$ ” variables:  $q_1, \dots, q_n$

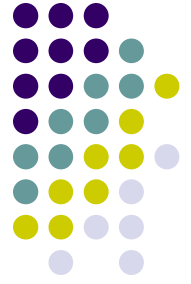
$$f : \mathbb{R}^n \rightarrow \mathbb{R}$$

- However, each of these variables  $q_i$  in turn depends on a set of “ $k$ ” other variables  $x_1, \dots, x_k$ .

$$\mathbf{q} = \begin{bmatrix} q_1(x_1, \dots, x_k) \\ \dots \\ q_n(x_1, \dots, x_k) \end{bmatrix} : \mathbb{R}^k \rightarrow \mathbb{R}^n$$

- The composition of  $f$  and  $\mathbf{q}$  leads to a new function  $\phi(\mathbf{x})$ :

$$\phi(\mathbf{x}) = f \circ \mathbf{q} = f(\mathbf{q}(\mathbf{x})) : \mathbb{R}^k \rightarrow \mathbb{R}$$



# Chain Rule for a Scalar Function

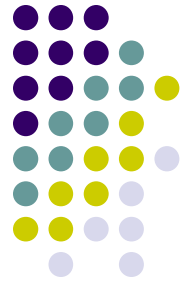
- The question: how do you compute  $\phi_{\mathbf{x}}$  ?
  - Using our notation:

$$\phi = f \circ \mathbf{q} = f(\mathbf{q}(\mathbf{x})) \quad \Rightarrow \quad \phi_{\mathbf{x}} = \frac{\partial \phi}{\partial \mathbf{x}} = ??$$

- Chain Rule for scalar function:

$$\phi_{\mathbf{x}} = \frac{\partial \phi}{\partial \mathbf{x}} = \frac{\partial f}{\partial \mathbf{q}} \cdot \frac{\partial \mathbf{q}}{\partial \mathbf{x}} = f_{\mathbf{q}} \cdot \mathbf{q}_{\mathbf{x}}$$

# Example



Assume that  $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$  and a function  $\phi$  of  $\mathbf{y}$  is defined as:  $\phi(\mathbf{y}) = 3y_1^2 + \sin y_2$ .

In turn,  $\mathbf{y}$  depends on a variable  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}$  as follows:

$$\mathbf{y} = \mathbf{y}(\mathbf{x}) = \begin{bmatrix} y_1(\mathbf{x}) \\ y_2(\mathbf{x}) \end{bmatrix} = \begin{bmatrix} 2x_1 + \log x_2 + \sqrt{x_3} \\ (x_1 - x_2)^2 \end{bmatrix}$$

Now, since  $\phi$  depends on  $\mathbf{y}$  and  $\mathbf{y}$  depends on  $\mathbf{x}$ , it means that  $\phi$  depends on  $\mathbf{x}$ . Find the partial derivative of  $\phi$  with respect to  $\mathbf{x}$ , that is,

$$\phi_{\mathbf{x}} = \frac{\partial \phi}{\partial \mathbf{x}} = \left[ \frac{\partial \phi}{\partial x_1} \quad \frac{\partial \phi}{\partial x_2} \quad \frac{\partial \phi}{\partial x_3} \right] = ?$$

## Scenario 2: Vector Function



- $\mathbf{F}$  is a function of “n” variables:  $q_1, \dots, q_n$   
$$\mathbf{F} : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

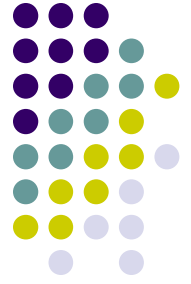
- However, each of these variables  $q_i$  in turn depends on a set of “k” other variables  $x_1, \dots, x_k$ .

$$\mathbf{q} = \begin{bmatrix} q_1(x_1, \dots, x_k) \\ \dots \\ q_n(x_1, \dots, x_k) \end{bmatrix} : \mathbb{R}^k \rightarrow \mathbb{R}^n$$

- The composition of  $\mathbf{F}$  and  $\mathbf{q}$  leads to a new function  $\Phi(\mathbf{x})$ :

$$\Phi(\mathbf{x}) = \mathbf{F} \circ \mathbf{q} = \mathbf{F}(\mathbf{q}(\mathbf{x})) : \mathbb{R}^k \rightarrow \mathbb{R}^m$$

# Chain Rule for a Vector Function



- How do you compute the partial derivative of  $\Phi$ ?

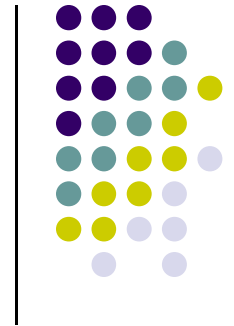
$$\Phi : \mathbb{R}^k \rightarrow \mathbb{R}^m$$

$$\Phi = \Phi(\mathbf{q}(\mathbf{x})) \quad \Rightarrow \quad \Phi_{\mathbf{x}} = \frac{\partial \Phi}{\partial \mathbf{x}} = ??$$

- Chain rule for vector functions:

$$\Phi_{\mathbf{x}} = \frac{\partial \Phi}{\partial \mathbf{x}} = \frac{\partial \Phi}{\partial \mathbf{q}} \cdot \frac{\partial \mathbf{q}}{\partial \mathbf{x}}$$

# Example



Assume that  $\mathbf{y} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$  and a function  $\mathbf{f}$  of  $\mathbf{y}$  is defined as:  $\mathbf{f}(\mathbf{y}) = \begin{bmatrix} 2y_1 + y_2^2 \\ y_1y_2 \end{bmatrix}$ .

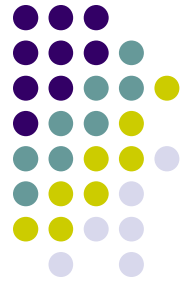
In turn,  $\mathbf{y}$  depends on a variable  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$  as follows:

$$\mathbf{y} = \mathbf{y}(\mathbf{x}) = \begin{bmatrix} y_1(\mathbf{x}) \\ y_2(\mathbf{x}) \end{bmatrix} = \begin{bmatrix} x_1x_2 \\ x_1^2 - x_2 \end{bmatrix}$$

Now, since  $\mathbf{f}$  depends on  $\mathbf{y}$  and  $\mathbf{y}$  depends on  $\mathbf{x}$ , it means that  $\mathbf{f}$  depends on  $\mathbf{x}$ . Find the partial derivative of  $\mathbf{f}$  with respect to  $\mathbf{x}$ , that is,

$$\mathbf{f}_{\mathbf{x}} = \frac{\partial \mathbf{f}}{\partial \mathbf{x}} = \left[ \frac{\partial \mathbf{f}}{\partial x_1} \quad \frac{\partial \mathbf{f}}{\partial x_2} \right] = ?$$

# Scenario 3: Function of Two Vectors



- $\mathbf{F}$  is a vector function of 2 vector variables  $\mathbf{q}$  and  $\mathbf{p}$ :

$$\mathbf{F} : \mathbb{R}^n \rightarrow \mathbb{R}^m$$

- Both  $\mathbf{q}$  and  $\mathbf{p}$  in turn depend on a set of “k” other variables  $\mathbf{x} = [x_1, \dots, x_k]^T$ :

$$\mathbf{q} = \mathbf{q}(x_1, \dots, x_k) : \mathbb{R}^k \rightarrow \mathbb{R}^{n_1}$$

$$\mathbf{p} = \mathbf{p}(x_1, \dots, x_k) : \mathbb{R}^k \rightarrow \mathbb{R}^{n_2}$$

$$n = n_1 + n_2$$

- A new function  $\Phi(\mathbf{x})$  is defined as:

$$\Phi(\mathbf{x}) = \mathbf{F}(\mathbf{q}(\mathbf{x}), \mathbf{p}(\mathbf{x})) : \mathbb{R}^k \rightarrow \mathbb{R}^m$$

# The Chain Rule



- How do you compute the partial derivative of  $\Phi$  with respect to  $\mathbf{x}$  ?

$$\Phi : \mathbb{R}^k \rightarrow \mathbb{R}^m$$

$$\Phi = \Phi(\mathbf{x}) \quad \Rightarrow \quad \Phi_{\mathbf{x}} = \frac{\partial \Phi}{\partial \mathbf{x}} = ??$$

- Chain rule for function of two vectors:

$$\Phi_{\mathbf{x}} = \frac{\partial \Phi}{\partial \mathbf{x}} = \frac{\partial \mathbf{F}}{\partial \mathbf{q}} \cdot \frac{\partial \mathbf{q}}{\partial \mathbf{x}} + \frac{\partial \mathbf{F}}{\partial \mathbf{p}} \cdot \frac{\partial \mathbf{p}}{\partial \mathbf{x}} = \mathbf{F}_{\mathbf{q}} \cdot \mathbf{q}_{\mathbf{x}} + \mathbf{F}_{\mathbf{p}} \cdot \mathbf{p}_{\mathbf{x}}.$$



# Example:



Assume that  $\mathbf{q} = \mathbf{q}(\mathbf{x}) \in \mathbb{R}^3$ , and  $\mathbf{p} = \mathbf{p}(\mathbf{x}) \in \mathbb{R}^3$ . Show that:

$$\frac{\partial(\mathbf{q}^T \mathbf{p})}{\partial \mathbf{x}} = \mathbf{q}^T \mathbf{p}_{\mathbf{x}} + \mathbf{p}^T \mathbf{q}_{\mathbf{x}}$$

# Scenario 4: Time Derivatives

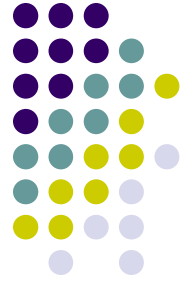


- In the previous slides we talked about functions  $f$  of  $y$ , while  $y$  in turn depended on yet another variable  $x$
- The relevant case is when the variable  $x$  is actually time,  $t$ 
  - This scenario is super common in 451:
    - You have a function that depends on the generalized coordinates  $\mathbf{q}$ , and in turn the generalized coordinates are functions of time (they change in time, since we are talking about kinematics/dynamics here...)
  - Case 1: scalar function that depends on an array of  $m$  generalized coordinates that in turn depend on time

$$\Phi = \Phi(\mathbf{q}(t)) \in \mathbb{R}$$

- Case 2: vector function (of dimension  $n$ ) that depends on an array of  $m$  generalized coordinates that in turn depend on time

$$\Phi = \Phi(\mathbf{q}(t)) \in \mathbb{R}^n$$



## A Special Case: Time Derivatives (Cntd)

- Of interest if finding the time derivative of  $\Phi$  and  $\dot{\Phi}$
- Apply the chain rule, the scalar function  $\Phi$  case first:

$$\dot{\Phi} = \frac{d\Phi}{dt} = \frac{d\Phi(\mathbf{q}(t))}{dt} = \frac{\partial\Phi}{\partial\mathbf{q}} \cdot \frac{d\mathbf{q}}{dt} = \Phi_{\mathbf{q}}\dot{\mathbf{q}} \in \mathbb{R}$$

- For the vector function case, applying the chain rule leads to the same formula, only the size of the result is different...

$$\dot{\Phi} = \frac{d\Phi}{dt} = \frac{d\Phi(\mathbf{q}(t))}{dt} = \frac{\partial\Phi}{\partial\mathbf{q}} \cdot \frac{d\mathbf{q}}{dt} = \Phi_{\mathbf{q}}\dot{\mathbf{q}} \in \mathbb{R}^n$$

# Example, Scalar Function $\Phi$



- Assume a set of generalized coordinates is defined through array  $\mathbf{q}$ . Also, a scalar function  $\Phi$  of  $\mathbf{q}$  is provided:

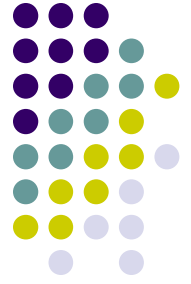
$$\mathbf{q}(t) = \begin{bmatrix} x(t) \\ y(t) \\ \theta(t) \end{bmatrix}$$

$$\Phi(\mathbf{q}) = 3x(t) + 2L \sin \theta(t)$$

- Find time derivative of  $\Phi$

$$\dot{\Phi} = ?$$

# Example, Vector Function $\Phi$



- Assume a set of generalized coordinates is defined through array  $\mathbf{q}$ . Also, a vector function  $\Phi$  of  $\mathbf{q}$  is provided:

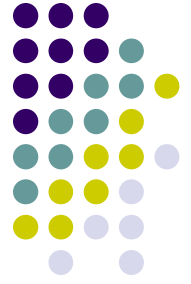
$$\mathbf{q}(t) = \begin{bmatrix} x(t) \\ y(t) \\ \theta(t) \end{bmatrix}$$

$$\Phi(\mathbf{q}) = \begin{bmatrix} 3x(t) & + & 2L \sin \theta(t) \\ y(t) & - & 2L \cos \theta(t) \end{bmatrix}$$

- Find time derivative of  $\Phi$

$$\dot{\Phi} = ?$$

# Useful Formulas



- A couple of useful formulas, some of them you had to derive as part of the HW

$$\frac{\partial(\mathbf{g}^T \mathbf{p})}{\partial \mathbf{q}} = \mathbf{g}^T \mathbf{p}_{\mathbf{q}} + \mathbf{p}^T \mathbf{g}_{\mathbf{q}}$$

$$\frac{\partial}{\partial \mathbf{q}} (\mathbf{C}\mathbf{q}) = \mathbf{C}$$

$$\frac{\partial}{\partial \mathbf{x}} (\mathbf{x}^T \mathbf{C}\mathbf{y}) = \mathbf{y}^T \mathbf{C}^T$$

$$\frac{d}{dt} (\mathbf{p}^T \mathbf{C}\mathbf{q}) = \dot{\mathbf{p}}^T \mathbf{C}\mathbf{q} + \mathbf{p}^T \mathbf{C}\dot{\mathbf{q}}$$

Assumptions:

$$\mathbf{g} = \mathbf{g}(\mathbf{q})$$

$$\mathbf{p} = \mathbf{p}(\mathbf{q})$$

$\mathbf{C}$  - constant matrix

$\mathbf{y}$  doesn't depend on  $\mathbf{x}$

The dimensions of the vectors and matrix above such that all the operations listed can be carried out.

# Example

- Derive the last equality on previous slide
- Can you expand that equation further?

$$\frac{d}{dt}(\mathbf{p}^T \mathbf{C} \mathbf{q}) = \dot{\mathbf{p}}^T \mathbf{C} \mathbf{q} + \mathbf{p}^T \mathbf{C} \dot{\mathbf{q}}$$

Assumptions:  
 $\mathbf{p} = \mathbf{p}(\mathbf{q})$   
 $\mathbf{C}$  - constant matrix

