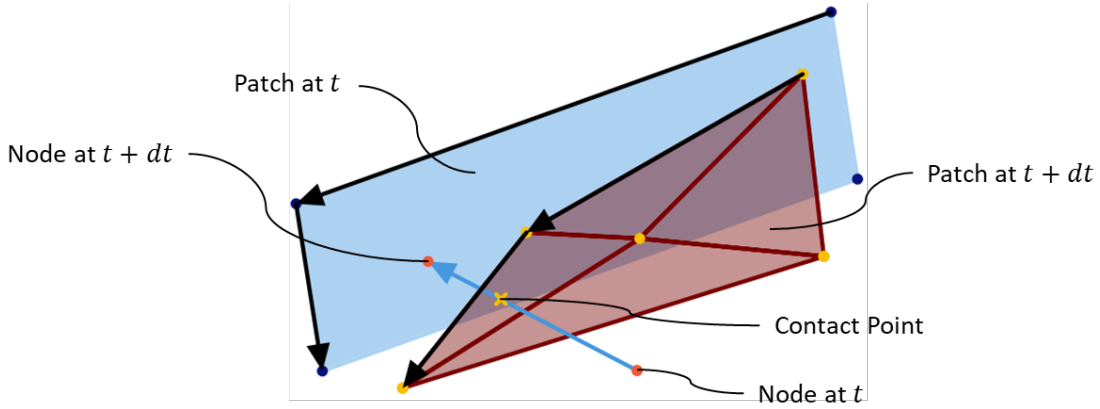


Contact Point to Reference

March 13, 2024

```
[1]: import sympy as sp
import numpy as np

xc, yc = sp.symbols('x_c y_c') # x and y values of contact point
xi, eta = sp.symbols(r'\xi eta') # reference coordinate variables
p, n = sp.symbols('p n') # p is the index of an array and n is the total
    ↳number of points
xp, yp = sp.Function('x'), sp.Function('y') # x and y components of a vector
xi_p, eta_p = sp.Function(r'\xi'), sp.Function(r'\eta') # xi and eta points
    ↳that make up the boundary of the surface (either 1 or -1 always)
```



This demo is provided for constructing the set of non-linear functions to solve for the ξ and η reference coordinates of the contact point and construct a Newton-Raphson scheme.

For mapping a reference point (ξ, η) to the global/actual position point (\vec{s}) , we use the following

$$\vec{s} = \sum_{p=0}^{n-1} \phi_p(\xi, \eta) \vec{s}_p$$

where $\phi_p(\xi, \eta) = \frac{1}{4}(1 + \xi_p \xi)(1 + \eta_p \eta)$ is the basis/shape function for 2D corresponding to a known reference point \vec{s}_p . The position point has components

$$\vec{s}_p = \begin{bmatrix} x_p \\ y_p \end{bmatrix}$$

At some contact point (ξ_c, η_c) , we can set up the following equation below to be analyzed.

```
[2]: phi_p = (sp.Rational(1, 4)*(1 + xi*xi_p(p))*(1 + eta*eta_p(p))).simplify() #
      ↪shape function in 2D space
phi_p_func = sp.Function(r'\phi_p')(xi, eta, p)
s_p = sp.Matrix([xp(p), yp(p)]) # Position vector of surface point
b = sp.Matrix([xc, yc])

eq = sp.Eq(b, sp.Sum(phi_p_func*s_p, (p, 0, n - 1)), evaluate=False)
display(eq)
eq1 = sp.Eq(b, sp.Sum(phi_p*s_p, (p, 0, n - 1)), evaluate=False)
display(eq1)
```

$$\begin{bmatrix} x_c \\ y_c \end{bmatrix} = \sum_{p=0}^{n-1} \begin{bmatrix} \phi_p(\xi, \eta, p)x(p) \\ \phi_p(\xi, \eta, p)y(p) \end{bmatrix}$$

$$\begin{bmatrix} x_c \\ y_c \end{bmatrix} = \sum_{p=0}^{n-1} \begin{bmatrix} \frac{(\eta\eta(p)+1)(\xi\xi(p)+1)x(p)}{4} \\ \frac{(\eta\eta(p)+1)(\xi\xi(p)+1)y(p)}{4} \end{bmatrix}$$

The $x(p)$, $\xi(p)$, and so on should be interpreted as x_p , ξ_p , and so on. This is how we can use sympy to symbolically construct the Newton-Raphson scheme in terms of reference points. For the Newton-Raphson scheme, we have

$$\begin{bmatrix} \xi_{i+1} \\ \eta_{i+1} \end{bmatrix} = \begin{bmatrix} \xi_i \\ \eta_i \end{bmatrix} - \mathbf{J}^{-1}\mathbf{F}$$

```
[3]: # Constructing the vector function F
F = eq.rhs.doit() - eq.lhs
F
```

```
[3]:
```

$$\begin{bmatrix} -x_c + \sum_{p=0}^{n-1} \phi_p(\xi, \eta, p)x(p) \\ -y_c + \sum_{p=0}^{n-1} \phi_p(\xi, \eta, p)y(p) \end{bmatrix}$$

```
[4]: # Constructing the jacobian J
jac = F.jacobian([xi, eta])
jac
```

```
[4]:
```

$$\begin{bmatrix} \sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p) & \sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p) \\ \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p) & \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p) \end{bmatrix}$$

```
[5]: # Construct the inverse jacobian J^{-1}
jac_inv = jac.inv()
jac_inv
```

```
[5]:
```

$$\begin{bmatrix} \frac{\sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)}{-\left(\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)\right) \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p) + \left(\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p)\right) \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)} & -\frac{\sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)}{-\left(\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)\right) \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p) + \left(\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p)\right) \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)} \\ -\frac{\sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p)}{-\left(\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)\right) \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p) + \left(\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p)\right) \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)} & -\frac{\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p)}{-\left(\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)\right) \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p) + \left(\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p)\right) \sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)} \end{bmatrix}$$

```
[6]: # The denominator of each item is messy
# To clean it up, I'm replacing it with a variable "d"
a11 = jac_inv[0, 0]
_, d = sp.fraction(a11)
jac_inv.subs(d, sp.Symbol("d"))
```

$$[6]: \begin{bmatrix} \frac{\sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)}{d} & -\frac{\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \eta} \phi_p(\xi, \eta, p)}{d} \\ -\frac{\sum_{p=0}^{n-1} y(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p)}{d} & \frac{\sum_{p=0}^{n-1} x(p) \frac{\partial}{\partial \xi} \phi_p(\xi, \eta, p)}{d} \end{bmatrix}$$

In summary, we have

$$F = \begin{bmatrix} -x_c + \sum_{p=0}^{n-1} \phi_p(\xi, \eta) x_p \\ -y_c + \sum_{p=0}^{n-1} \phi_p(\xi, \eta) y_p \end{bmatrix}$$

$$J = \begin{bmatrix} \sum_{p=0}^{n-1} x_p \frac{\partial}{\partial \xi} \phi_p(\xi, \eta) & \sum_{p=0}^{n-1} x_p \frac{\partial}{\partial \eta} \phi_p(\xi, \eta) \\ \sum_{p=0}^{n-1} y_p \frac{\partial}{\partial \xi} \phi_p(\xi, \eta) & \sum_{p=0}^{n-1} y_p \frac{\partial}{\partial \eta} \phi_p(\xi, \eta) \end{bmatrix}$$

$$J^{-1} = \begin{bmatrix} \frac{\sum_{p=0}^{n-1} y_p \frac{\partial}{\partial \eta} \phi_p(\xi, \eta)}{d} & -\frac{\sum_{p=0}^{n-1} x_p \frac{\partial}{\partial \eta} \phi_p(\xi, \eta)}{d} \\ -\frac{\sum_{p=0}^{n-1} y_p \frac{\partial}{\partial \xi} \phi_p(\xi, \eta)}{d} & \frac{\sum_{p=0}^{n-1} x_p \frac{\partial}{\partial \xi} \phi_p(\xi, \eta)}{d} \end{bmatrix}$$

where

$$d = - \left(\sum_{p=0}^{n-1} x_p \frac{\partial}{\partial \eta} \phi_p(\xi, \eta) \right) \sum_{p=0}^{n-1} y_p \frac{\partial}{\partial \xi} \phi_p(\xi, \eta) + \left(\sum_{p=0}^{n-1} x_p \frac{\partial}{\partial \xi} \phi_p(\xi, \eta) \right) \sum_{p=0}^{n-1} y_p \frac{\partial}{\partial \eta} \phi_p(\xi, \eta)$$

The code below is the numerical implementation of this scheme.

```
[7]: #TODO: Implement the phi_p_lamb function and also its derivative
class Node:
    def __init__(self, pos, ref_pos):
        """
        :param pos: Physical position array
        :param ref_pos: Reference position array
        """
        self.x, self.y, self.z = pos
        self.xi, self.eta, self.zeta = ref_pos

    def phi_p_lamb(xi_, eta_, xi_p_, eta_p_):
        return 0.25*(1 + xi_*xi_p_)*(1 + eta_*eta_p_)

    def get_F(reference_point, physical_point, nodes):
        xi_, eta_ = reference_point
        xc_, yc_ = physical_point
        return np.array([
```

```

        0.25*sum([eta_*xi_*p_.eta*p_.xi*p_.x + eta*p_.eta*p_.x + xi_*p_.xi*p_.
↪x + p_.x for p_ in nodes]) - xc_,
        0.25*sum([eta_*xi_*p_.eta*p_.xi*p_.y + eta*p_.eta*p_.y + xi_*p_.xi*p_.
↪y + p_.y for p_ in nodes]) - yc_
    ])

def get_jacobian_inverse(reference_point, nodes):
    xi_, eta_ = reference_point
    den = (sum([eta*p_.eta*p_.xi*p_.x + p_.xi*p_.x for p_ in nodes])/
↪4*sum([xi_*p_.eta*p_.xi*p_.y + p_.eta*p_.y for p_ in nodes])/4 -
        sum([eta*p_.eta*p_.xi*p_.y + p_.xi*p_.y for p_ in nodes])/
↪4*sum([xi_*p_.eta*p_.xi*p_.x + p_.eta*p_.x for p_ in nodes])/4)
    return np.array([
        [sum([xi_*p_.eta*p_.xi*p_.y + p_.eta*p_.y for p_ in nodes])/4/den,
↪-sum([xi_*p_.eta*p_.xi*p_.x + p_.eta*p_.x for p_ in nodes])/4/den],
        [-sum([eta*p_.eta*p_.xi*p_.y + p_.xi*p_.y for p_ in nodes])/4/den,
↪sum([eta*p_.eta*p_.xi*p_.x + p_.xi*p_.x for p_ in nodes])/4/den]
    ])

def newton_raphson(reference_point, physical_point, nodes, tol=1e-8,
↪max_iter=100):
    xi_, eta_ = reference_point
    for i in range(max_iter):
        F_ = get_F([xi_, eta_], physical_point, nodes)
        jac_inv_ = get_jacobian_inverse([xi_, eta_], nodes)
        xi_, eta_ = np.array([xi_, eta_]) - jac_inv_ @ F_
        if np.linalg.norm(F_) < tol:
            break
    # noinspection PyUnboundLocalVariable
    return np.array([xi_, eta_]), i

```

Consider a quadrilateral surface bound by the following points:

Label	ξ, η, ζ	x, y, z
0	-1, -1, -1	0.51025339, 0.50683559, 0.99572776
1	1, -1, -1	1.17943427, 0.69225101, 1.93591633
2	1, 1, -1	0.99487331, 0.99743665, 2.97094874
3	-1, 1, -1	0.49444608, 0.99700943, 1.96411315

The contact point is (0.92088978, 0.74145551, 1.89717136). The analysis omits ζ because we already know that the contact point is on the exterior surface. For this case, $\zeta = -1$. **Note: The implemented procedure needs to use those reference points that are changing.** For example, if contact is on the reference plane $\eta = 1$, then the process needs to solve for ξ and ζ .

```

[8]: patch_nodes = [
    Node([0.51025339, 0.50683559, 0.99572776], [-1, -1, -1]),

```

```
Node([1.17943427, 0.69225101, 1.93591633], [1, -1, -1]),  
Node([0.99487331, 0.99743665, 2.97094874], [1, 1, -1]),  
Node([0.49444608, 0.99700943, 1.96411315], [-1, 1, -1])  
]  
  
newton_raphson([0.5, -0.5], [0.92088978, 0.74145551], patch_nodes)
```

```
[8]: (array([ 0.34340497, -0.39835547]), 3)
```