#### **Two-Dimensional Diffusion**

The governing balance equations that describe diffusion processes in situations involving two independent variables appear typically as

$$\nabla \cdot (k \nabla u(x, y, t)) + \rho \frac{\partial u}{\partial t} + f(x, y, t) = 0$$
 in  $\Omega$ 

$$u = g(s, t)$$
 on  $\Gamma_1$ 

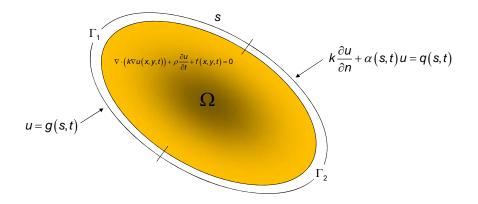
$$k \frac{\partial u}{\partial n} + \alpha(s, t) u = q(s, t)$$
 on  $\Gamma_2$ 

$$u(x, y, 0) = c(x, y)$$
 in  $\Omega$ 

#### TIME-DEPENDENT PROBLEMS

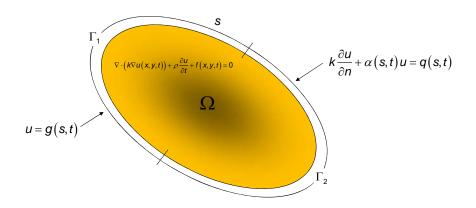
#### **Two-Dimensional Diffusion**

Where  $\Omega$  is the interior domain, and  $\Gamma_{\rm 1}$  and  $\Gamma_{\rm 2}$  form the boundary of the domain.



#### **Two-Dimensional Diffusion**

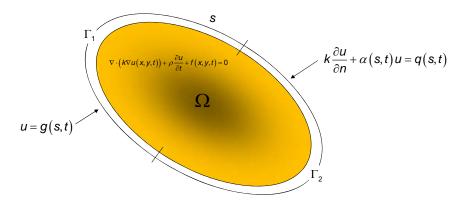
The physical constants are *k* and  $\rho$ , and *s* is a linear measure of the position on the boundary.



#### TIME-DEPENDENT PROBLEMS

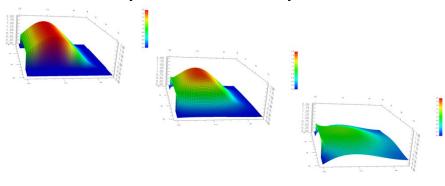
#### **Two-Dimensional Diffusion**

The type of boundary condition specified on  $\Gamma_2$  results from a local balance between conduction in the interior and convection into the exterior.



#### **Two-Dimensional Diffusion**

In two dimensions, the influence of the boundary conditions is shown. The left side of the simulation domain (x = 0) is used as a constant source of diffusion particles, thereby creating a steady flow into the area. All other boundaries are modeled by a Neumann boundary condition.



#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

- This equation is similar in form to the elliptic boundary value problem studied in Chapter 3, with the very important addition of the time derivative term in the differential equation and the corresponding initial condition.
- With these additions, the problem changes from an elliptic boundary value problem to a parabolic initial boundary value problem.

#### **Two-Dimensional Diffusion**

- The basic steps of discretization, interpolation, elemental formulation, assembly, constraints, solution, and computation of derived variables are presented in this section as they relate to the two-dimensional parabolic initial-boundary value problem.
- The Galerkin method, in connection with the corresponding weak formulation to be developed, will be used to generate the finite element model.
- **Discretization** For discretization, please refer to the material in Chapter 3.

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

- **Interpolation -** The solution is assumed to be expressible in terms of the nodally based interpolation functions  $n_i(x, y)$  introduced and discussed in Section 3.2.
- In the present setting, these interpolation functions are used with the semidiscretization

$$u(x, y, t) = \sum_{1}^{N+1} u_i(t) n_i(x, y)$$

The  $n_i(x, y)$  are nodally based interpolation functions and can be linear, quadratic, or as otherwise desired.

#### **Two-Dimensional Diffusion**

- The Galerkin Finite Element Method
- **Elemental Formulation -** The starting point for the elemental formulation is the weak formulation of the initial-boundary value problem.
- The first step in developing the weak formulation is to multiply the differential equation by an arbitrary test function v(x, y) vanishing on  $\Gamma_1$ .

The result is then integrated over the domain  $\Omega$  to obtain

$$\iint_{\Omega} v \left( \nabla \cdot \left( k \nabla u \right) + \rho \frac{\partial u}{\partial t} + f \right) d\Omega = 0$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Elemental Formulation -** Using the two-dimensional form of the divergence theorem to integrate the first term by parts, there results after rearranging

$$\iint_{\Omega} \nabla v \cdot k \nabla u \, d\Omega + \iint_{\Omega} v \rho \frac{\partial u}{\partial t} \, d\Omega$$
$$= \int_{\Gamma} v \mathbf{n} \cdot k \nabla u \, d\Gamma + \iint_{\Omega} v f \, d\Omega$$

#### where $\Gamma = \Gamma_1 + \Gamma_2$

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Elemental Formulation** - Recalling that *v* vanishes on  $\Gamma_1$ and that  $k\partial u/\partial x = k\mathbf{n} \cdot \nabla u = q - hu$  on  $\Gamma_2$ , it follows that

$$\iint_{\Omega} \nabla v \cdot k \nabla u \, d\Omega + \iint_{\Omega} v \rho \frac{\partial u}{\partial t} \, d\Omega$$
$$= \int_{\Gamma} v (q - hu) \, d\Gamma + \iint_{\Omega} v f \, d\Omega$$

This equation is the required weak formulation for the twodimensional diffusion problem.

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Elemental Formulation -** Substituting the approximation of u(x, y, t) into the weak formulation and taking  $v = n_k$ , k = 1, 2, ... yields

$$\iint_{\Omega^{-}} \left( \frac{\partial n_{k}}{\partial x} k \sum u_{i} \frac{\partial n_{i}}{\partial x} + \frac{\partial n_{k}}{\partial y} k \sum u_{i} \frac{\partial n_{i}}{\partial y} \right) d\Omega$$
$$+ \iint_{\Omega^{-}} n_{k} \rho \sum \dot{u}_{i} n_{i} d\Omega + \int_{\Gamma^{-}} n_{k} h \sum u_{i} n_{i} d\Gamma$$
$$= \iint_{\Omega^{-}} n_{k} f d\Omega + \int_{\Gamma^{-}_{2}} n_{k} q d\Gamma \qquad k = 1, 2, ...$$

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Elemental Formulation –** Where  $\Omega^{-}$  and  $\Gamma_{2}^{-}$  represent the elemental areas approximating  $\Omega$ , and the collection of the elemental edges approximating  $\Gamma_{2}$ , respectively.

$$\iint_{\Omega^{-}} \left( \frac{\partial n_{k}}{\partial x} k \sum u_{i} \frac{\partial n_{i}}{\partial x} + \frac{\partial n_{k}}{\partial y} k \sum u_{i} \frac{\partial n_{i}}{\partial y} \right) d\Omega$$
$$+ \iint_{\Omega^{-}} n_{k} \rho \sum \dot{u}_{i} n_{i} d\Omega + \int_{\Gamma^{-}} n_{k} h \sum u_{i} n_{i} d\Gamma$$
$$= \iint_{\Omega^{-}} n_{k} f d\Omega + \int_{\Gamma^{-}_{2}} n_{k} q d\Gamma \qquad k = 1, 2, ...$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Elemental Formulation –** This *N* x *N* set of linear algebraic equations can be written as:

$$\sum_{i=1}^{N} \left( A_{ki} u_{i} + B_{ki} \dot{u}_{i} \right) = F_{k}(t) \qquad k = 1, 2, ..., N$$
$$A_{ki} = \iint_{\Omega^{-}} \left( \frac{\partial n_{k}}{\partial x} k \frac{\partial n_{i}}{\partial x} + \frac{\partial n_{k}}{\partial y} k \frac{\partial n_{i}}{\partial y} \right) d\Omega + \int_{\Gamma^{-}} n_{k} h n_{i} d\Gamma$$
$$B_{ki} = \iint_{\Omega^{-}} n_{k} \rho n_{i} d\Omega$$
$$F_{k} = \iint_{\Omega^{-}} n_{k} f d\Omega + \int_{\Gamma^{-}_{2}} n_{k} q d\Gamma$$

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Elemental Formulation** – Note that assembly is contained implicitly within the formulation.

$$\sum_{i=1}^{N} \left( A_{ki} u_{i} + B_{ki} \dot{u}_{i} \right) = F_{k}(t) \qquad k = 1, 2, ..., N$$
$$A_{ki} = \iint_{\Omega^{-}} \left( \frac{\partial n_{k}}{\partial x} k \frac{\partial n_{i}}{\partial x} + \frac{\partial n_{k}}{\partial y} k \frac{\partial n_{i}}{\partial y} \right) d\Omega + \int_{\Gamma^{-}} n_{k} h n_{i} d\Gamma$$
$$B_{ki} = \iint_{\Omega^{-}} n_{k} \rho n_{i} d\Omega$$
$$F_{k} = \iint_{\Omega^{-}} n_{k} f d\Omega + \int_{\Gamma^{-}_{2}} n_{k} q d\Gamma$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Elemental Formulation** – In terms of the corresponding elementally based interpolations

$$U_e(x, y) = \mathbf{N}^T \mathbf{u}_e = \mathbf{u}_e^T \mathbf{N}$$

The finite element model can be expressed as

#### $\bm{A}\bm{u}+\bm{B}\dot{\bm{u}}=\bm{F}$

$$\mathbf{A} = \sum_{e} \mathbf{k}_{G} + \sum_{e} \mathbf{a}_{G} \qquad \mathbf{B} = \sum_{e} \mathbf{r}_{G}$$
$$\mathbf{F} = \sum_{e} \mathbf{f}_{G} + \sum_{e} \mathbf{q}_{G}$$

#### **Two-Dimensional Diffusion**

The Galerkin Finite Element Method

Elemental Formulation – In terms of the corresponding elementally based interpolations

$$u_e(x, y) = \mathbf{N}^T \mathbf{u}_e = \mathbf{u}_e^T \mathbf{N}$$

The finite element model can be expressed as

$$\mathbf{k}_{e} = \iint_{A_{e}} \left( \frac{\partial \mathbf{N}}{\partial x} k \frac{\partial \mathbf{N}^{T}}{\partial x} + \frac{\partial \mathbf{N}}{\partial y} k \frac{\partial \mathbf{N}^{T}}{\partial y} \right) dA$$
$$\mathbf{r}_{e} = \iint_{A_{e}} \mathbf{N} \rho \mathbf{N}^{T} dA \qquad \mathbf{a}_{e} = \int_{\gamma_{2e}} \mathbf{N} \alpha \mathbf{N}^{T} dS$$
$$\mathbf{f}_{e} = \iint_{A_{e}} \mathbf{N} f dA \qquad \mathbf{q}_{e} = \int_{\gamma_{2e}} \mathbf{N} q dS$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Elemental Formulation –** The initial conditions for the system of first-order differential equations are obtained from the initial conditions prescribed for the original initial-boundary value problem.

Generally  $\mathbf{u}(0)$  is determined by evaluating the function c(x, y) at the nodes to obtain

$$\mathbf{u}(0) = \mathbf{u}_0 = \left\langle \boldsymbol{c}_1 \quad \boldsymbol{c}_2 \quad \boldsymbol{c}_3 \quad \dots \quad \boldsymbol{c}_{N-1} \quad \boldsymbol{c}_N \right\rangle^{\mathsf{T}}$$

where  $c_i = c(x_i, y_i)$  with  $(x_i, y_i)$  the coordinates of the *i*<sup>th</sup> node.

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Constraints** - The constraints arise from the boundary conditions specified on  $\Gamma_1$ .

- Generally the values of the constraints are determined from the *q* function with the constrained value of *u* at a node on  $\Gamma_1$  being taken as the value of *q* at that point.
- These constraints are then enforced on the assembled equations, resulting in the final global constrained set of linear first-order differential equations.

$$\mathbf{M}\dot{\mathbf{u}} + \mathbf{K}\mathbf{u} = \mathbf{f}$$
  $\mathbf{u}(0) = \mathbf{u}_0$ 

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

- **Solution** The system of equations is precisely the same in form and character as the corresponding equations developed for one-dimensional diffusion.
- An analytical method as well as the numerical methods of Euler and improved Euler or Crank-Nicolson can be used for integrating the above set of equations.

#### **Two-Dimensional Diffusion**

#### The Galerkin Finite Element Method

**Solution** - The Euler method will be conditionally stable with the critical time step depending on the maximum eigenvalue of the associated problem  $(\mathbf{K} - \lambda \mathbf{M})\mathbf{v} = 0$ .

The improved Euler or Crank-Nicolson algorithm will be unconditionally stable.

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

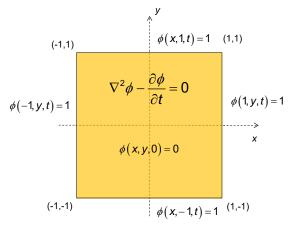
#### The Galerkin Finite Element Method

**Derived variables** - The derived variables will be time dependent, and depending on the particular problem being considered, may need to be computed at each time step.

The computations would be per element and would be carried out using the techniques described in Chapter 3.

#### **Two-Dimensional Diffusion - Example 1**

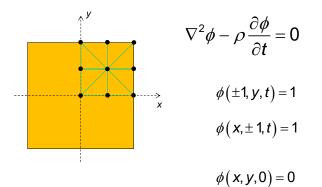
Consider the **Two-Dimensional Diffusion - Example 1** problem shown below.



#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

**Discretization.** Due to symmetry, we will model the top right-most quadrant of the membrane using eight equally-sized 3-node triangles.



#### **Two-Dimensional Diffusion - Example 1**

**Elemental Formulation** – This *N* x *N* set of linear algebraic equations can be written as:

$$\sum_{i=1}^{N} \left( A_{ki} \phi_{i} + B_{ki} \dot{\phi}_{i} \right) = 0 \qquad k = 1, 2, \dots, N$$
$$A_{ki} = \iint_{\Omega^{-}} \left( \frac{\partial n_{k}}{\partial x} k \frac{\partial n_{i}}{\partial x} + \frac{\partial n_{k}}{\partial y} k \frac{\partial n_{i}}{\partial y} \right) d\Omega$$
$$B_{ki} = \iint_{\Omega^{-}} n_{k} \rho n_{i} d\Omega$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

**Elemental Formulation** – In terms of the corresponding elementally based interpolations

$$\phi_e(\mathbf{X},\mathbf{Y}) = \mathbf{N}^T \phi_e = \phi_e^T \mathbf{N}$$

The finite element model can be expressed as

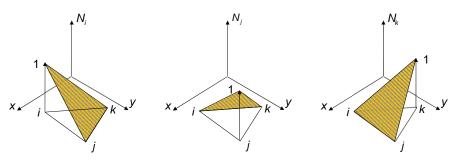
$$\mathbf{k}_{e} = \iint_{A_{e}} \left( \frac{\partial \mathbf{N}}{\partial x} k \frac{\partial \mathbf{N}^{T}}{\partial x} + \frac{\partial \mathbf{N}}{\partial y} k \frac{\partial \mathbf{N}^{T}}{\partial y} \right) dA$$
$$\mathbf{r}_{e} = \iint_{A_{e}} \mathbf{N} \rho \mathbf{N}^{T} dA$$

#### **Two-Dimensional Diffusion - Example 1**

**Interpolation** - In matrix notation, the distribution of the function over the element is:

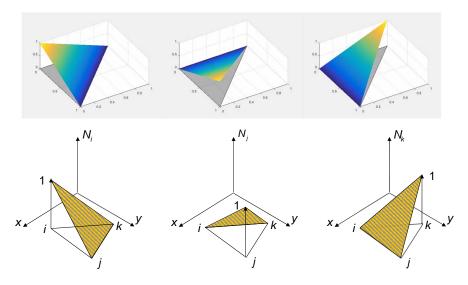
$$\phi_{\rm e}(\mathbf{X}, \mathbf{Y}) = \phi_{\rm e}^{\rm T} \mathbf{N} = \mathbf{N}^{\rm T} \phi_{\rm e}$$

The linear triangular shape functions are illustrated below:



#### TIME-DEPENDENT PROBLEMS

**Two-Dimensional Diffusion - Example 1** 



#### **Two-Dimensional Diffusion - Example 1**

**Interpolation** - The derivatives of *u* over the element with respect to both coordinates are:

$\frac{\partial \phi_{e}(x,y)}{\partial x} = \phi_{e}^{T}$	∂N	∂N <sup>™</sup>	$\frac{\partial \phi_{e}(x, y)}{\partial y} = \phi_{e}^{T}$	∂N _	∂N <sup>™</sup>
$\partial x = \psi_{e}$	$\partial x$	$= \frac{\partial x}{\partial x} \varphi_{e}$	$\partial y = \varphi_{e}$	∂y _	∂y <sup>− ϕ</sup> e

Calculating the derivatives of the shape functions gives:

$\frac{\partial \mathbf{N}}{\partial x} = \frac{\mathbf{b}_{e}}{2A_{e}}$	$\frac{\partial \mathbf{N}}{\partial y} = \frac{\mathbf{c}_{e}}{2A_{e}}$
$\mathbf{b}_{\mathbf{e}}^{T} = \left\langle b_{i}  b_{j}  b_{k} \right\rangle$	$\mathbf{C}_{\mathbf{e}}^{\mathbf{T}} = \left\langle \mathbf{C}_{i}  \mathbf{C}_{j}  \mathbf{C}_{k} \right\rangle$
$\boldsymbol{b}_i = \boldsymbol{y}_j - \boldsymbol{y}_k$	$m{c}_i = m{x}_k - m{x}_j$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

- **Elemental Formulation** The integrals defined in **k**<sub>e</sub> are the elemental "stiffness" matrix.
- For the linear triangular element we have discussed the stiffness matrix reduces to:

$$\mathbf{k}_{\mathbf{e}} = \iint_{A_{\mathbf{e}}} \left[ \frac{\mathbf{b}_{\mathbf{e}} \mathbf{b}_{\mathbf{e}}^{\mathsf{T}} + \mathbf{c}_{\mathbf{e}} \mathbf{c}_{\mathbf{e}}^{\mathsf{T}}}{4A_{\mathbf{e}}^{2}} \right] dA$$

Since the integrand of **k**<sub>e</sub> is a constant, the elemental stiffness matrix becomes:

$$\mathbf{k}_{\mathbf{e}} = \frac{\mathbf{b}_{\mathbf{e}} \mathbf{b}_{\mathbf{e}}^{\mathsf{T}} + \mathbf{c}_{\mathbf{e}} \mathbf{c}_{\mathbf{e}}^{\mathsf{T}}}{4A_{\mathbf{e}}}$$

The resulting is a 3x3 elemental stiffness matrix.

#### **Two-Dimensional Diffusion - Example 1**

The elemental **k**<sub>e</sub> matrix using 3-noded triangular elements is:

$$\mathbf{k}_{\mathbf{e}} = \frac{\mathbf{b}_{\mathbf{e}} \mathbf{b}_{\mathbf{e}}^{\mathsf{T}} + \mathbf{c}_{\mathbf{e}} \mathbf{c}_{\mathbf{e}}^{\mathsf{T}}}{4A_{\mathbf{e}}}$$
$$\mathbf{b}_{\mathbf{e}} = \begin{cases} \mathbf{y}_{j} - \mathbf{y}_{k} \\ \mathbf{y}_{k} - \mathbf{y}_{i} \\ \mathbf{y}_{i} - \mathbf{y}_{j} \end{cases} \qquad \mathbf{c}_{\mathbf{e}} = \begin{cases} \mathbf{x}_{k} - \mathbf{x}_{j} \\ \mathbf{x}_{i} - \mathbf{x}_{k} \\ \mathbf{x}_{j} - \mathbf{x}_{i} \end{cases}$$

#### TIME-DEPENDENT PROBLEMS

**Two-Dimensional Diffusion - Example 1** For element 1: node 1 = (0, 0); node 3 = (1/2, 1/2); and node 4 = (0, 1/2).  $\mathbf{b}_1 = \frac{1}{2} \begin{cases} 0\\1\\-1 \end{cases}$   $\mathbf{c}_1 = \frac{1}{2} \begin{cases} -1\\0\\1 \end{bmatrix}$   $A_1 = \frac{1}{8}$  $\mathbf{k}_1 = \frac{1}{2} \begin{bmatrix} 1 & 0 & -1\\0 & 1 & -1\\-1 & -1 & 2 \end{bmatrix} \begin{bmatrix} 1\\0\\1 \end{bmatrix}$ 



#### For element 2:

node 1 = (0, 0); node 2 =  $(\frac{1}{2}, 0)$ ; and node 3 =  $(\frac{1}{2}, \frac{1}{2})$ .

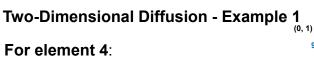
$$\mathbf{b}_2 = \frac{1}{2} \begin{cases} -1 \\ 1 \\ 0 \end{cases} \qquad \mathbf{c}_2 = \frac{1}{2} \begin{cases} 0 \\ -1 \\ 1 \end{cases}$$

 $\mathbf{k}_{2} = \frac{1}{2} \begin{bmatrix} 1 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \\ 3 \end{bmatrix}$ 

Die 1  
(0, 1) (<sup>1</sup>/<sub>2</sub>, 1) (1, 1)  
d (0, <sup>1</sup>/<sub>2</sub>) (<sup>1</sup>/<sub>2</sub>, <sup>1</sup>/<sub>2</sub>) (<sup>1</sup>/<sub>2</sub>, <sup>1</sup>/<sub>2</sub>)  
(0, <sup>1</sup>/<sub>2</sub>) (<sup>1</sup>/<sub>2</sub>, <sup>1</sup>/<sub>2</sub>) (<sup>1</sup>/<sub>2</sub>, <sup>1</sup>/<sub>2</sub>)  
(0, 0) (<sup>1</sup>/<sub>2</sub>, <sup>1</sup>/<sub>2</sub>) (<sup>1</sup>/<sub>2</sub>, <sup>1</sup>/<sub>2</sub>) (<sup>1</sup>/<sub>2</sub>, <sup>1</sup>/<sub>2</sub>)  
= 
$$\frac{1}{8}$$

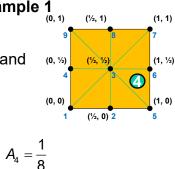
 $A_2$ 

#### TIME-DEPENDENT PROBLEMS

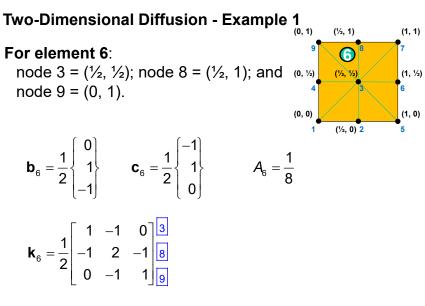


node 5 = (1, 0); node 6 = (1,  $\frac{1}{2}$ ); and node 3 = ( $\frac{1}{2}$ ,  $\frac{1}{2}$ ).

$$\mathbf{b}_{4} = \frac{1}{2} \begin{cases} 0\\1\\-1 \end{cases} \qquad \mathbf{c}_{4} = \frac{1}{2} \begin{cases} -1\\1\\0 \end{cases}$$
$$\mathbf{k}_{4} = \frac{1}{2} \begin{bmatrix} 1 & -1 & 0\\-1 & 2 & -1\\0 & -1 & 1 \end{bmatrix} \begin{bmatrix} 5\\6\\3 \end{bmatrix}$$

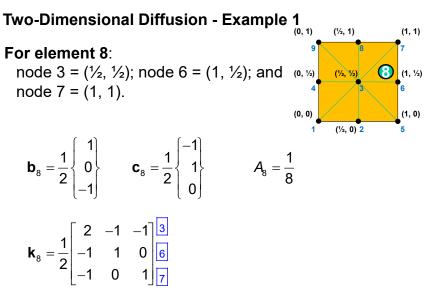


#### TIME-DEPENDENT PROBLEMS



#### TIME-DEPENDENT PROBLEMS

**Two-Dimensional Diffusion - Example 1** For element 7: node 3 =  $(\frac{1}{2}, \frac{1}{2})$ ; node 7 = (1, 1); and node 8 =  $(\frac{1}{2}, 1)$ .  $\mathbf{b}_7 = \frac{1}{2} \begin{cases} 0\\1\\-1 \end{cases}$   $\mathbf{c}_7 = \frac{1}{2} \begin{cases} -1\\0\\1 \end{bmatrix}$   $\mathbf{c}_7 = \frac{1}{2} \begin{cases} -1\\0\\1 \end{bmatrix}$   $A_7 = \frac{1}{8}$  $\mathbf{k}_7 = \frac{1}{2} \begin{bmatrix} 1 & 0 & -1\\0 & 1 & -1\\-1 & -1 & 2 \end{bmatrix} \begin{bmatrix} 3\\7\\8 \end{bmatrix}$ 



#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

The elemental  $\mathbf{m}_{e}$  matrices are:

$$\mathbf{m}_{e} = \iint_{A_{e}} \mathbf{N} \mathbf{N}^{\mathsf{T}} \, dA = \iint_{A_{e}} \begin{cases} N_{1} \\ N_{2} \\ N_{3} \end{cases} \langle N_{1} \quad N_{2} \quad N_{3} \rangle \, dA$$
$$= \frac{A_{e}}{12} \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix} \qquad A_{e} = \frac{1}{8}$$
$$\mathbf{m}_{1} = \mathbf{m}_{2} = \dots = \mathbf{m}_{8} = \frac{1}{96} \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix}$$

#### **Two-Dimensional Diffusion - Example 1**

The assembled  $\mathbf{K}_{\mathbf{G}}$  and  $\mathbf{M}_{\mathbf{G}}$  matrices are:

	2	-1	0	-1	0	0	0	0	0]	$\phi_1$
$\mathbf{K}_{\mathbf{G}} = \frac{1}{2}$	-1	4	-2	0	-1	0	0	0	0	$\phi_2$
	0	-2	8	-2	0	-2	0	-2	0	$\phi_3$
	-1	0	-2	4	0	0	0	0	-1	$\phi_4$
$\mathbf{K}_{\mathbf{G}} = \frac{1}{2}$	0	-1	0	0	2	-1	0	0	0	$\phi_{5}$
2	0	0	-2	0	-1	4	-1	0	0	$\phi_{6}$
	0	0	0	0	0	-1	2	-1	0	<b>\$\$</b> _7
	0	0	-2 0	0	0	0	-1	4	-1	$\phi_8$
	0	0	0	-1	0	0	0	-1	2	<b>ø</b> 9

#### TIME-DEPENDENT PROBLEMS

**Two-Dimensional Diffusion - Example 1** 

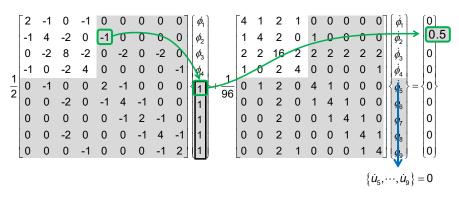
The assembled  $\mathbf{K}_{\mathbf{G}}$  and  $\mathbf{M}_{\mathbf{G}}$  matrices are:

$$\mathbf{M}_{\mathbf{G}} = \frac{1}{96} \begin{bmatrix} 4 & 1 & 2 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 4 & 2 & 0 & 1 & 0 & 0 & 0 & 0 \\ 2 & 2 & 16 & 2 & 2 & 2 & 2 & 2 & 2 \\ 1 & 0 & 2 & 4 & 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 2 & 0 & 4 & 1 & 0 & 0 & 0 \\ 0 & 0 & 2 & 0 & 1 & 4 & 1 & 0 & 0 \\ 0 & 0 & 2 & 0 & 0 & 1 & 4 & 1 & 0 \\ 0 & 0 & 2 & 0 & 0 & 0 & 1 & 4 & 1 \\ 0 & 0 & 2 & 1 & 0 & 0 & 0 & 1 & 4 \end{bmatrix} \begin{pmatrix} \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \\ \phi_5 \\ \phi_6 \\ \phi_7 \\ \phi_8 \\ \phi_9 \end{bmatrix}$$

#### **Two-Dimensional Diffusion - Example 1**

Applying nonhomogeneous boundary conditions that  $\{\phi_5, \dots, \phi_9\} = \{1, 1, 1, 1, 1\}$ ; the constrained **K**<sub>G</sub> is:

$$\mathbf{K}_{\mathbf{G}}\phi + \mathbf{M}_{\mathbf{G}}\dot{\phi} = \mathbf{0}$$



#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

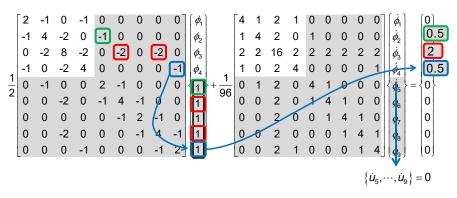
Applying nonhomogeneous boundary conditions that  $\{\phi_5, \dots, \phi_9\} = \{1, 1, 1, 1, 1\}$ ; the constrained **K**<sub>G</sub> is:

$$\mathbf{K}_{\mathbf{G}}\phi + \mathbf{M}_{\mathbf{G}}\dot{\phi} = \mathbf{0}$$

#### **Two-Dimensional Diffusion - Example 1**

Applying nonhomogeneous boundary conditions that  $\{\phi_5, \dots, \phi_9\} = \{1, 1, 1, 1, 1\}$ ; the constrained **K**<sub>G</sub> is:

$$\mathbf{K}_{\mathbf{G}}\phi + \mathbf{M}_{\mathbf{G}}\dot{\phi} = \mathbf{0}$$



#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

**Assembly.** With both the boundary conditions essential, **BT**=0 and **bt**=0. It follows that the assembled equations are:

$$\mathbf{K}_{\mathbf{G}}\phi + \mathbf{M}_{\mathbf{G}}\dot{\phi} = \mathbf{0}$$

	2	-1	0	-1]	$\left( \phi_{1} \right)$		4	1	2	1	$\left \left(\dot{\phi}_{1}\right)\right $		(0)	
1	-1	4	-2	0	$\phi_2$	1	1	4	2	0	$\dot{\phi}_2$	_ 1	1	
2	0	-2	8	-2	$\phi_3$	<sup>+</sup> 96	2	2	16	2	$\dot{\phi}_3$	$\frac{1}{2}$	4	ĺ
	1	0	-2	4	$\phi_4$	+ 1/96	1	0	2	4	$\dot{\phi}_4$		[ 1]	ļ

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\theta\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\theta)\mathbf{K}\right)\phi_{n}+h\left[\theta\mathbf{f}_{n+1}+(1-\theta)\mathbf{f}_{n}\right]$$

It is easily seen that:  $\begin{cases} \theta = 0 & \text{Euler method} \\ \theta = \frac{1}{2} & \text{Crank-Nicolson method} \end{cases}$ 

The value  $\theta$  = 1 corresponds to what is referred to as the modified Euler method and corresponds to using a backward difference scheme obtained by evaluating the differential equation at  $t_{n+1}$  and taking:

$$\dot{\phi}_{n+1} = \frac{\phi_{n+1} - \phi_n}{h}$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

For the  $\theta$  algorithm, recall the largest eigenvalue of the **K** –  $\lambda$ **M** help define the time step:

$$h < \frac{2}{\lambda_{\max}}$$

The eigenvalue problem is:

$$\left(\mathbf{K}_{\mathbf{G}} + \lambda \mathbf{M}_{\mathbf{G}}\right)\phi = \mathbf{0}$$

$$\begin{pmatrix} 2 & -1 & 0 & -1 \\ -1 & 4 & -2 & 0 \\ 0 & -2 & 8 & -2 \\ -1 & 0 & -2 & 4 \end{bmatrix} - \frac{\lambda}{96} \begin{bmatrix} 4 & 1 & 2 & 1 \\ 1 & 4 & 2 & 0 \\ 2 & 2 & 16 & 2 \\ 1 & 0 & 2 & 4 \end{bmatrix} \phi = 0$$

#### **Two-Dimensional Diffusion - Example 1**

The eigenvalues can be found using a variety of available solution techniques. In Matlab, use [V,D] = eig(A).

For example,  $[V,D] = eig(M_{G}^{-1}K_{G})$  gives eigenvalues D:

 $\lambda_1 = 5.6265$   $\lambda_2 = 32$   $\lambda_3 = 48$   $\lambda_4 = 102.3735$ 

The eigenvectors v are:

<b>V</b> <sub>1</sub> = <	(-0.69829)		0.89443		0.00000		(-0.44393)	
	-0.44187	V -	0.00000	> V <sub>3</sub> = {	-0.70711	<i>V</i> <sub>4</sub> = -	0.61386	
	-0.34914	$\mathbf{v}_2 = \mathbf{v}_2$	-0.44721		0.00000		-0.22197	Ì
	_0.44187		0.00000		0.70711		0.61386	J

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

For the example the largest eigenvalue is 102.3735, the critical time step is:

$$h < \frac{2}{\lambda_{\max}} < \frac{2}{102.3735} < 0.0195 \, \text{sec}.$$

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+(1-\boldsymbol{\theta})\mathbf{f}_{n}\right]$$

In this example, let's assume h = 0.001 sec. and  $\theta = \frac{1}{2}$ , therefore:

$$\left(\mathbf{M} + \mathbf{h}\boldsymbol{\theta}\mathbf{K}\right) = \frac{1}{96} \begin{bmatrix} 4 & 1 & 2 & 1 \\ 1 & 4 & 2 & 0 \\ 2 & 2 & 16 & 2 \\ 1 & 0 & 2 & 4 \end{bmatrix} + \begin{pmatrix} 0.001 \end{pmatrix} \frac{1}{2} \begin{pmatrix} 1 \\ 2 \end{pmatrix} \begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 4 & -2 & 0 \\ 0 & -2 & 8 & -2 \\ -1 & 0 & -2 & 4 \end{bmatrix}$$
$$= \begin{bmatrix} 0.04217 & 0.01017 & 0.02083 & 0.01017 \\ 0.01017 & 0.04267 & 0.02033 & 0.00000 \\ 0.02083 & 0.02033 & 0.16867 & 0.02033 \\ 0.01017 & 0.00000 & 0.02033 & 0.04267 \end{bmatrix}$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

In this example, let's assume h = 0.001 sec. and  $\theta = \frac{1}{2}$ , therefore:

 $\left(\mathbf{M} + \mathbf{h}\mathbf{\theta}\mathbf{K}\right)^{-1} = \begin{bmatrix} 27.40488 & -5.55522 & -2.04559 & -5.55522 \\ -5.55522 & 26.08491 & -2.77761 & 2.64741 \\ -2.04559 & -2.77761 & 6.85122 & -2.77761 \\ -5.55522 & 2.64741 & -2.77761 & 26.08491 \end{bmatrix}$ 

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

In this example, let's assume h = 0.001 sec. and  $\theta = \frac{1}{2}$ , therefore:

$$\left(\mathbf{M} - \mathbf{h}\boldsymbol{\theta}\mathbf{K}\right) = \frac{1}{96} \begin{bmatrix} 4 & 1 & 2 & 1 \\ 1 & 4 & 2 & 0 \\ 2 & 2 & 16 & 2 \\ 1 & 0 & 2 & 4 \end{bmatrix} - \left(0.001\right) \frac{1}{2} \left(\frac{1}{2}\right) \begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 4 & -2 & 0 \\ 0 & -2 & 8 & -2 \\ -1 & 0 & -2 & 4 \end{bmatrix}$$
$$= \begin{bmatrix} 0.04117 & 0.01067 & 0.02083 & 0.01067 \\ 0.01067 & 0.04067 & 0.02133 & 0.00000 \\ 0.02083 & 0.02133 & 0.16467 & 0.02133 \\ 0.01067 & 0.00000 & 0.02133 & 0.04067 \end{bmatrix}$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

In this example, let's assume h = 0.001 sec. and  $\theta = \frac{1}{2}$ , therefore:

$$\mathbf{f} = \frac{1}{2} \begin{cases} 0 \\ 1 \\ 4 \\ 1 \end{cases} \implies h \Big[ \theta \mathbf{f}_{n+1} + (1 - \theta) \mathbf{f}_n \Big] = \frac{(0.001)}{2} \begin{cases} 0 \\ 1 \\ 4 \\ 1 \end{cases} = \begin{cases} 0.00000 \\ 0.00050 \\ 0.00050 \\ 0.00050 \end{cases}$$

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

In this example, let's assume h = 0.001 sec. and  $\theta = \frac{1}{2}$ , therefore:

$$\mathbf{u}_{n+1} = \left(\mathbf{M} + h\theta\mathbf{K}\right)^{-1} \left[ \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\mathbf{u}_{n} + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_{n}\right] \right]$$

$$\begin{cases} \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{4} \\ \phi_{4} \\ \phi_{4} \\ \phi_{1} \\ \phi_{4} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{1} \\ \phi_{1} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2} \\ \phi_{2} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2} \\ \phi_{2} \\ \phi_{2} \\ \phi_{1} \\ \phi_{2} \\ \phi_{2}$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+(1-\boldsymbol{\theta})\mathbf{f}_{n}\right]$$

For n = 1, with u(x, y, 0) = 0, then:

$$\begin{cases} \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{4} \\ \phi_{3} \\ \phi_{4} \\ \phi_$$

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+\left(1-\boldsymbol{\theta}\right)\mathbf{f}_{n}\right]$$

For n = 2, with u(x, y, 0) = 0, then:

$\left( \phi_{1} \right)$		0.96704	0.02277	-0.00293	0.02277	(-0.00965)		(-0.00965)
$\phi_2$		0.01992	0.94227	0.03984	-0.01085	0.00881		0.00881
$\phi_3$	> =	-0.00073	0.01138	0.96704	0.01138	0.01092	> + <	0.01092
$\left[\phi_{4}\right]$	2	0.01992	-0.01085	0.03984	0.94227	0.00881	1	0.00881
$ \begin{bmatrix} \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \end{bmatrix} $	2	(-0.01861) 0.01726 0.02170 0.01726		$\dot{\phi}_2 = \frac{\phi_2}{h}$	$\frac{-\phi_1}{1} = \begin{cases} 8.4\\ 10.7 \end{cases}$	5924 4984 77241 4984		

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+\left(1-\boldsymbol{\theta}\right)\mathbf{f}_{n}\right]$$

For n = 3, with u(x, y, 0) = 0, then:

$$\begin{cases} \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \\ \phi_3 \\ \phi_4 \\ \phi_3 \\ \phi_4 \\ \phi_3 \\ \phi_4 \\ \phi_3 \\ \phi_4 \\ \phi_$$

#### **Two-Dimensional Diffusion - Example 1**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+\left(1-\boldsymbol{\theta}\right)\mathbf{f}_{n}\right]$$

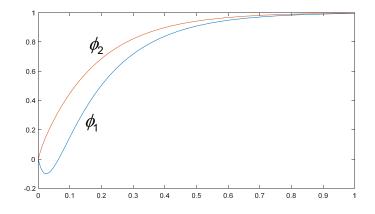
For n = 100, with u(x, y, 0) = 0, then:

$\left[\phi_{1}\right]$		0.96704	0.02277	-0.00293	0.02277	(0.14298)		-0.00965
$\phi_2$	_	0.01992	0.94227	0.03984	-0.01085	0.44394		0.00881
$\phi_3$		-0.00073	0.01138	0.96704	0.01138	0.54977	> + <	0.01092
$\phi_4$	100	0.01992	-0.01085	0.03984	0.94227	0.44394	99	0.00881
$\left( \phi_{1} \right)$		(0.14722)				(4.24609)	)	
$\phi_2$	_	0.44706		$\dot{\phi}_{10}$	$\phi_{99}$ _	3.11999		
$\phi_3$	• =	0.55257		$\phi_{100} =$	h	2.80721	ĺ	
$\phi_4$	100	0.44706				3.11999	ļ	

#### TIME-DEPENDENT PROBLEMS

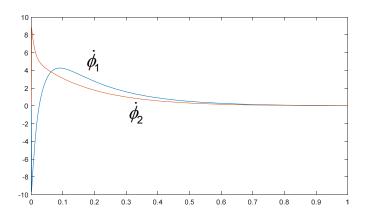
#### **Two-Dimensional Diffusion - Example 2**

The values of  $\phi_1$  and  $\phi_2$  for  $0 \le t \le 1$  sec. are show below:

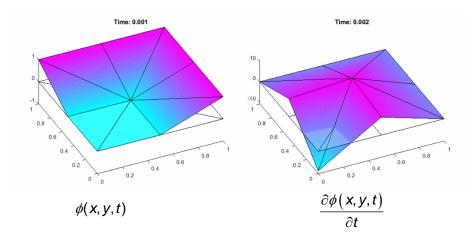


#### **Two-Dimensional Diffusion - Example 2**

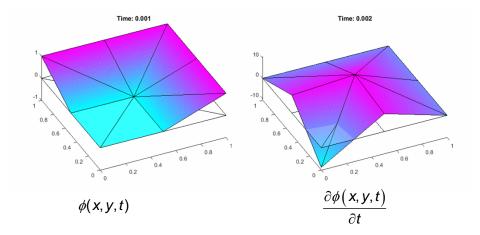
The derivative values of  $\phi_1$  and  $\phi_2$  for 0 < t < 1 sec. are:



## TIME-DEPENDENT PROBLEMS Two-Dimensional Diffusion - Example 1



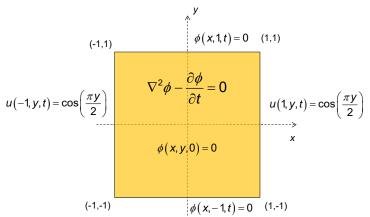
**Two-Dimensional Diffusion - Example 1** 



#### TIME-DEPENDENT PROBLEMS

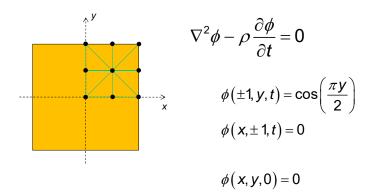
#### **Two-Dimensional Diffusion - Example 2**

Consider the two-dimensional diffusion problem shown below.



#### **Two-Dimensional Diffusion - Example 2**

**Discretization.** Due to symmetry, we will model the top right-most quadrant of the membrane using eight equally-sized 3-node triangles.



#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

Applying homogeneous boundary conditions that  $u_7$ ,  $u_8$ , and  $u_9 = 0$ ; then the assembled **K**<sub>G</sub> and **M**<sub>G</sub> matrices are:

$$\mathbf{K}_{\mathbf{G}} = \frac{1}{2} \begin{bmatrix} 2 & -1 & 0 & -1 & 0 & 0 \\ -1 & 4 & -2 & 0 & -1 & 0 \\ 0 & -2 & 8 & -2 & 0 & -2 \\ -1 & 0 & -2 & 4 & 0 & 0 \\ 0 & -1 & 0 & 0 & 2 & -1 \\ 0 & 0 & -2 & 0 & -1 & 4 \end{bmatrix} \begin{pmatrix} \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \\ \phi_5 \\ \phi_6 \end{bmatrix}$$

#### **Two-Dimensional Diffusion - Example 2**

Applying homogeneous boundary conditions that  $u_7$ ,  $u_8$ , and  $u_9 = 0$ ; then the assembled **K**<sub>G</sub> and **M**<sub>G</sub> matrices are:

		4	1	2	1	0	0	$\dot{\phi}_1$
М	<u>1</u> 96	1	4	2	0	1	0	$\dot{\phi}_2$
		2	2	16	2	2	2	$\dot{\phi}_3$
IVI <sub>G</sub> —		1	0	2	4	0	0	$\dot{\phi}_4$
		0	1	2	0	4	1	$\dot{\phi}_5$
		0	0	2	0	1	4	$\dot{\phi}_6$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

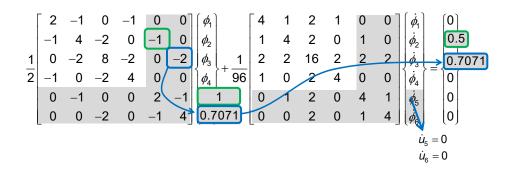
**Assembly.** With both the boundary conditions essential, **BT**=0 and **bt**=0. It follows that the assembled equations are:

$$K_{G}u + M_{G}\dot{u} = 0$$
  $u_{5} = 1$  and  $u_{6} = 0.7071$ 

#### **Two-Dimensional Diffusion - Example 2**

**Assembly.** With both the boundary conditions essential, **BT**=0 and **bt**=0. It follows that the assembled equations are:

$$\mathbf{K}_{\mathbf{G}}\mathbf{u} + \mathbf{M}_{\mathbf{G}}\dot{\mathbf{u}} = 0$$
  $u_5 = 1$  and  $u_6 = 0.7071$ 



#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

**Assembly.** With both the boundary conditions essential, **BT**=0 and **bt**=0. It follows that the assembled equations are:

$$\mathbf{K}_{\mathbf{G}}\phi + \mathbf{M}_{\mathbf{G}}\dot{\phi} = \mathbf{0}$$

$$\frac{1}{2}\begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 4 & -2 & 0 \\ 0 & -2 & 8 & -2 \\ -1 & 0 & -2 & 4 \end{bmatrix} \begin{bmatrix} \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \end{bmatrix} + \frac{1}{96}\begin{bmatrix} 4 & 1 & 2 & 1 \\ 1 & 4 & 2 & 0 \\ 2 & 2 & 16 & 2 \\ 1 & 0 & 2 & 4 \end{bmatrix} \begin{bmatrix} \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 0 \\ 1 \\ \sqrt{2} \\ 0 \end{bmatrix}$$

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+(1-\boldsymbol{\theta})\mathbf{f}_{n}\right]$$

It is easily seen that:  $\begin{cases} \theta = 0 & \text{Euler method} \\ \theta = \frac{1}{2} & \dot{\phi} & \text{Crank-Nicolson method} \end{cases}$ 

The value  $\theta$  = 1 corresponds to what is referred to as the *modified* Euler method and corresponds to using a **backward difference scheme** obtained by evaluating the differential equation at  $t_{n+1}$  and taking:

$$\dot{\phi}_{n+1} = \frac{\phi_{n+1} - \phi_n}{h}$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

In this example, let's assume h = 0.002 sec. and  $\theta = \frac{1}{2}$ , therefore:

$$\left(\mathbf{M} + h\boldsymbol{\theta}\mathbf{K}\right) = \frac{1}{96} \begin{bmatrix} 4 & 1 & 2 & 1^{\mathbf{\mu}} \\ 1 & 4 & 2 & 0 \\ 2 & 2 & 16 & 2 \\ 1 & 0 & 2 & 4 \end{bmatrix} + \begin{pmatrix} 0.002 \end{pmatrix} \frac{1}{2} \begin{pmatrix} 1 \\ 2 \end{pmatrix} \begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 4 & -2 & 0 \\ 0 & -2 & 8 & -2 \\ -1 & 0 & -2 & 4 \end{bmatrix}$$
$$= \begin{bmatrix} 0.04267 & 0.00992 & 0.02083 & 0.00992 \\ 0.00992 & 0.04367 & 0.01983 & 0.00000 \\ 0.02083 & 0.01983 & 0.17067 & 0.01983 \\ 0.00992 & 0.00000 & 0.01983 & 0.04367 \end{bmatrix}$$

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

In this example, let's assume h = 0.002 sec. and  $\theta = \frac{1}{2}$ , therefore:

 $\left(\mathbf{M} + \boldsymbol{h}\boldsymbol{\theta}\mathbf{K}\right)^{-1} = \begin{bmatrix} 26.84816 & -5.15256 & -2.07980 & -5.15256 \\ -5.15256 & 25.24105 & -2.57628 & 2.34028 \\ -2.07980 & -2.57628 & 6.71204 & -2.57628 \\ -5.15256 & 2.34028 & -2.57628 & 25.24105 \end{bmatrix}$ 

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

In this example, let's assume h = 0.002 sec. and  $\theta = \frac{1}{2}$ , therefore:

$$\left(\mathbf{M} - \mathbf{h}\boldsymbol{\theta}\mathbf{K}\right) = \frac{1}{96} \begin{bmatrix} 4 & 1 & 2 & 1 \\ 1 & 4 & 2 & 0 \\ 2 & 2 & 16 & 2 \\ 1 & 0 & 2 & 4 \end{bmatrix} - (0.002) \frac{1}{2} \left(\frac{1}{2}\right) \begin{bmatrix} 2 & -1 & 0 & -1 \\ -1 & 4 & -2 & 0 \\ 0 & -2 & 8 & -2 \\ -1 & 0 & -2 & 4 \end{bmatrix}$$
$$= \begin{bmatrix} 0.04067 & 0.01092 & 0.02083 & 0.01092 \\ 0.01092 & 0.03967 & 0.02183 & 0.00000 \\ 0.02083 & 0.02183 & 0.16267 & 0.02183 \\ 0.01092 & 0.00000 & 0.02183 & 0.03967 \end{bmatrix}$$

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#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

In this example, let's assume h = 0.002 sec. and  $\theta = \frac{1}{2}$ , therefore:

$$\mathbf{f} = \frac{1}{2} \begin{cases} 0 \\ 1 \\ \sqrt{2} \\ 0 \end{cases} \implies h \Big[ \theta \mathbf{f}_{n+1} + (1 - \theta) \mathbf{f}_n \Big] = \frac{(0.002)}{2} \begin{cases} 0 \\ 1 \\ \sqrt{2} \\ 0 \end{cases} = \begin{cases} 0.00000 \\ 0.00100 \\ 0.00141 \\ 0.00000 \end{cases}$$

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+(1-\boldsymbol{\theta})\mathbf{f}_{n}\right]$$

In this example, let's assume h = 0.002 sec. and  $\theta = \frac{1}{2}$ , therefore:

$$\mathbf{u}_{n+1} = \left(\mathbf{M} + h\theta\mathbf{K}\right)^{-1} \left[ \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\mathbf{u}_{n} + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_{n}\right] \right]$$

$$\begin{cases} \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{4} \\ \phi_{1+1} \end{cases} = \begin{bmatrix} 0.93600 & 0.04330 & -0.00397 & 0.04330 \\ 0.03789 & 0.88873 & 0.07577 & -0.01967 \\ -0.00099 & 0.02165 & 0.93600 & 0.02165 \\ 0.03789 & -0.01967 & 0.07577 & 0.88873 \end{bmatrix} \begin{bmatrix} \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi$$

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

For n = 1, with u(x, y, 0) = 0, then:

$\left(\phi_{1}\right)$		0.93600	0.04330	-0.00397	0.04330	[0]	[-0.00809]
$\phi_2$		0.03789	0.88873	0.07577	-0.01967	0	0.02160
$\phi_3$	} =	-0.00099	0.02165	0.93600	0.02165	0	+ 0.00692
$\phi_4$	$\int_{1}$	0.03789	-0.01967	0.07577	0.88873	[0] <sub>0</sub>	[-0.00130]
$\left[\phi_{1}\right]$		(-0.00809)		4	,	)4692	I
$\int \phi_2$	=	0.02160	,	$\dot{\phi}_{1} = \frac{\varphi_{1}}{\varphi_{1}}$	<u> </u>	79882	Ş
$\phi_3$		0.00692		$r_1 h$	3.4	15799	9
$\phi_4$	) <sub>1</sub>	[-0.00130]			(-0.6	65156	<b>5</b> ]

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M} + h\theta\mathbf{K}\right)\phi_{n+1} = \left(\mathbf{M} - h(1-\theta)\mathbf{K}\right)\phi_n + h\left[\theta\mathbf{f}_{n+1} + (1-\theta)\mathbf{f}_n\right]$$

For n = 2, with u(x, y, 0) = 0, then:

$$\begin{cases} \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \\ \phi_2 \\ \phi_3 \\ \phi_4 \\ \phi_2 \\ \phi_4 \\ \phi_2 \\ \phi_3 \\ \phi_4 \\ \phi_2 \\ \phi$$

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+\left(1-\boldsymbol{\theta}\right)\mathbf{f}_{n}\right]$$

For n = 3, with u(x, y, 0) = 0, then:

$\left( \phi_{1} \right)$		0.93600	0.04330	-0.00397	0.04330	(-0.01482)		(-0.00809)
$ \phi_2 $	_	0.03789	0.88873	0.07577	-0.01967	0.04104		0.02160
$\phi_3$	=	-0.00099	0.02165	0.93600	0.02165	0.01384	> + <	0.00692
$\left[\phi_{4}\right]_{3}$	3	0.03789	-0.01967	0.07577	0.88873	_0.00267	2	–0.00130
$\begin{bmatrix} \phi_1 \\ \phi_2 \\ \phi_3 \\ \phi_4 \end{bmatrix}_3$	= -	(-0.02036) 0.05861 0.02071 -0.00399	>	$\dot{\phi}_3 = \frac{\phi_3 - \phi_3}{h}$	$\frac{\phi_2}{\eta} = \begin{cases} 8.7\\ 3.4 \end{cases}$	76958 78560 13787 66308		

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

The  $\theta$  algorithm for the time integration is:

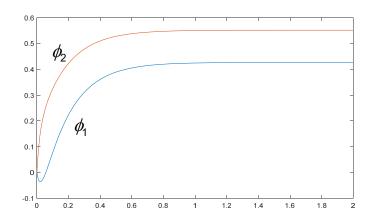
$$\left(\mathbf{M}+h\boldsymbol{\theta}\mathbf{K}\right)\phi_{n+1}=\left(\mathbf{M}-h(1-\boldsymbol{\theta})\mathbf{K}\right)\phi_{n}+h\left[\boldsymbol{\theta}\mathbf{f}_{n+1}+\left(1-\boldsymbol{\theta}\right)\mathbf{f}_{n}\right]$$

For n = 100, with u(x, y, 0) = 0, then:

$$\begin{cases} \phi_{1} \\ \phi_{2} \\ \phi_{3} \\ \phi_{4} \\ \phi_{100} \end{cases} = \begin{cases} 0.93600 & 0.04330 & -0.00397 & 0.04330 \\ 0.03789 & 0.88873 & 0.07577 & -0.01967 \\ -0.00099 & 0.02165 & 0.93600 & 0.02165 \\ 0.03789 & -0.01967 & 0.07577 & 0.88873 \end{bmatrix} \begin{cases} 0.21938 \\ 0.42032 \\ 0.28613 \\ 0.17034 \\ g_{99} \end{cases} + \begin{cases} -0.00809 \\ 0.02160 \\ 0.00692 \\ -0.00130 \end{cases}$$

#### **Two-Dimensional Diffusion - Example 2**

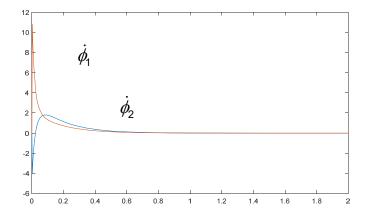
The values of  $\phi_1$  and  $\phi_2$  for 0 < t < 2 sec. are show below:



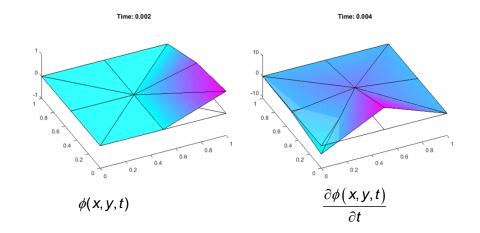
## TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion - Example 2**

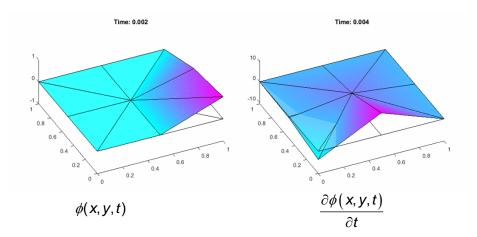
The derivative values of  $\phi_1$  and  $\phi_2$  for  $0 \le t \le 2$  sec. are:



#### **Two-Dimensional Diffusion - Example 2**



## TIME-DEPENDENT PROBLEMS Two-Dimensional Diffusion - Example 2



#### **Two-Dimensional Diffusion**

Time-dependent problems are inherently more difficult and expensive to solve than their corresponding steadystate counterparts.

The expense of generating the global matrices is higher for the time-dependent problems because of the necessity of computing the mass matrices.

The main extra expense, however, is in solving the resulting time-dependent global equations.

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

For an analytical approach to the solution, additional expense is incurred in terms of having to determine eigenvalues and eigenvectors.

The actual amount of expense depends on the specific form of the stiffness and mass matrices and the algorithm used, but in any case it is significantly in excess of the expense of solving the single set of linear algebraic equations associated with the steady-state problem.

#### **Two-Dimensional Diffusion**

- For a time domain integration technique, the additional expense is clearly related to the number of time steps necessary to trace out the desired time history.
- In addition to several matrix multiplications and additions, each step can involve the solution of a set of linear algebraic equations.
- In some instances this expense can be minimized by using a decomposition that can be reused for the computation of the solution at each new time.

#### TIME-DEPENDENT PROBLEMS

#### **Two-Dimensional Diffusion**

- In this regard recall that the Euler and central difference algorithms require that the size of the time step not exceed a value proportional to the inverse of the largest eigenvalue.
- For large systems this critical step size can be very small resulting in many applications of the algorithm to trace out the time history.
- The unconditionally stable Crank-Nicolson and Newmark algorithms, can be used with arbitrary step size that has been chosen so as to accurately integrate the lower modes, with significant improvement in the expense relative to the conditionally stable algorithms.

# End of Chapter 4c