

# Multi-level adaptive vertex-centered finite volume methods for diffusion problems

Fayssal BENKHALDOUN

supervising: Tarek Ghoudi - PhD

Joint work with Imad Kissami Postdoc

July 3, 2018



- Adaptive FE-FV are now widely used in the numerical solution of (PDEs) to achieve better accuracy with minimum degrees of freedom.
- We first solve the PDE to get the solution on the current mesh.
- The error is estimated using the solution, and used to mark a set of triangles that are to be refined.
- Triangles are refined in such a way to keep mesh regularity and conformity.



Modelization

Simulation

High Performance Calculation

- A typical loop of (AFE-FVM ) through local refinement involves:



- ✓ Conformity of the mesh
- ✓ Prevent the propagation of refinement levels
- ✓ Efficiency of estimator
- ✓ Convergence of error
- ✓ Performance of CPU time

Let  $a : \mathbb{R}_+ \rightarrow \mathbb{R}$  be a given nonlinear function.

Typically,  $a(x) = x^{p-2}$  for some real number  $p \in (1, +\infty)$ . Let  $\sigma$  such that

$$\sigma(\xi) = a(|\xi|)\xi \quad \forall \xi \in \mathbb{R}^d \quad (1)$$

where  $|\cdot|$  is the Euclidean norm in  $\mathbb{R}^d$ . Then, for a given source function  $f : \Omega \rightarrow \mathbb{R}$ , the nonlinear Laplace problem consists in looking for  $u : \Omega \rightarrow \mathbb{R}$  such that

$$\begin{aligned} -\operatorname{div}(\sigma(\nabla u)) &= f \text{ in } \Omega \\ u &= g \text{ on } \partial\Omega \end{aligned} \quad (2)$$

**Problem:** find  $p \in H_0^1(\Omega)$ ,  $(\mathcal{S}) \begin{cases} -\operatorname{div}(\mathbb{K}\nabla p) = f & \text{in } \Omega \subset \mathbb{R}^{d=2,3} \\ p = g & \text{on } \partial\Omega \end{cases}$  (3)

## Unicity and Existence

Assumptions:

- (H1)  $\mathbb{K} \in \mathbb{L}^\infty(\Omega)$ .
- (H2)  $f \in \mathbb{L}^2(\Omega)$ .
- $(\mathcal{S})$  has a unique solution.

**Remark:** The problem (2) represents, for instance, the extension of the problem (3) which takes into account the nonlinear dependence of the Darcy velocity on the pressure head gradient  $\nabla p$ . Note that (2) and (3) coincide, for  $a(x) = x^{p-2}$ , when  $p = 2$ .

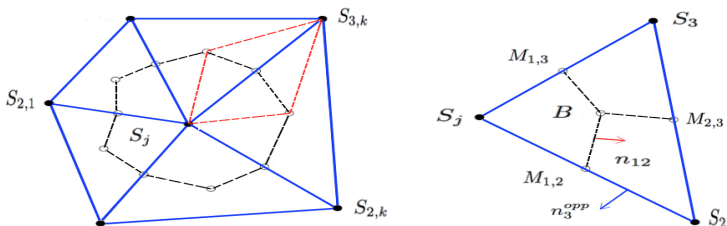


Figure: Dual cell

- $S_1$ ,  $S_2$ , and  $S_3$  are the vertices of a triangle  $T$ ,
- $B$  its barycentre,  $\Sigma_1^{opp}$ ,  $\Sigma_2^{opp}$  and  $\Sigma_3^{opp}$  The edges  $[S_2S_3]$ ,  $[S_1S_3]$  et  $[S_1S_2]$  ;
- $\vec{n}_1^{opp}$ ,  $\vec{n}_2^{opp}$  and  $\vec{n}_3^{opp}$  outgoing unit normals such that  $\vec{n}_{pq} \perp \overrightarrow{M_{pq}B}$  and  $\vec{n}_{pq} \cdot \overrightarrow{S_pS_q} > 0$

The approximation of the diffusive flux is based on an implicit scheme:

$$- \int_{\partial D_h} \mathbb{K} \nabla p \cdot \vec{n} \, d\sigma = \int_{D_h} f(x) \, dx \quad (4)$$

$$- \sum_{T \cap D_h \neq \emptyset} \int_{\partial D_h \cap T} \mathbb{K}_T \nabla p \cdot \vec{n} \, d\sigma = \int_{D_h} f(x) \, dx \quad (5)$$

We note the elementary diffusion terms by:

$$k_{12}^{flow}(T) = |T| \mathbb{K}_T \frac{|\Sigma_1^{opp}|}{2|T|} \frac{|\Sigma_2^{opp}|}{2|T|} \vec{n}_1^{opp} \vec{n}_2^{opp}$$

$$k_{13}^{flow}(T) = |T| \mathbb{K}_T \frac{|\Sigma_1^{opp}|}{2|T|} \frac{|\Sigma_3^{opp}|}{2|T|} \vec{n}_1^{opp} \vec{n}_3^{opp}$$

Finally, the finite volume scheme for the flow equation is written:

$$\sum_{T \in D_h} k_{12}^{flow}(T)(p_2 - p_1) + k_{13}^{flow}(T)(p_3 - p_1) = \int_{D_h} f(x) \, dx \quad (6)$$



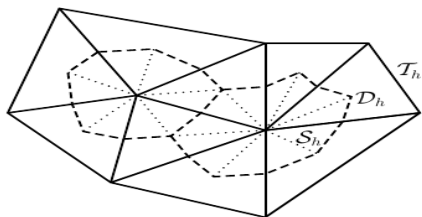


Figure: Primal mesh  $\mathcal{T}_h$ , Dual mesh  $\mathcal{D}_h$  and the fine simplicial mesh  $\mathcal{S}_h$

**Remark:** the flux  $-\mathbb{K}\nabla p \in H(\text{div}, \Omega)$  but  $-\mathbb{K}\nabla p_h \notin H(\text{div}, \Omega)$

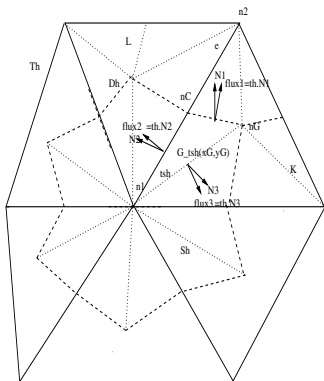
**Flux reconstruction** (exploits the local conservativity):

$$\mathbf{t}_h \in \text{RTN}_0(\mathcal{S}_h) \subset H(\text{div}, \Omega)$$

$$(\text{div } \mathbf{t}_h, 1)_D = (f, 1)_D, \quad \forall D \in \mathcal{D}_h^{\text{int}}$$

**Construction of  $\mathbf{t}_h$  by Direct Prescription** : We solved the following system ( $\mathcal{S}'$ ):

$$(\mathcal{S}') \begin{cases} \mathbf{t}_h \cdot \vec{N}_1 = -\mathbb{K} \nabla \mathbf{p}_h \cdot \vec{N}_1 \\ \mathbf{t}_h \cdot \vec{N}_2 = -w_{K,s} \left( \mathbb{K}_{|K} \nabla \mathbf{p}_h \cdot \vec{N}_2 \right) - w_{L,s} \left( \mathbb{K}_{|L} \nabla \mathbf{p}_h \cdot \vec{N}_2 \right) \\ \mathbf{t}_h \cdot \vec{N}_3 = -\mathbb{K} \nabla \mathbf{p}_h \cdot \vec{N}_3 \end{cases}$$



- $\mathbb{K}_{|K}$  ( $\mathbb{K}_{|L}$ ) is an approximation of the tensor of permeability on the triangle  $K$  ( $L$ )
- $\vec{N}_1$ ,  $\vec{N}_2$  and  $\vec{N}_3$  : unit normal vectors.
- Harmonic averaging :

$$w_{K,s} = \frac{\mathbb{K}_K}{\mathbb{K}_K + \mathbb{K}_L}, w_{L,s} = \frac{\mathbb{K}_L}{\mathbb{K}_K + \mathbb{K}_L}$$

## Error estimator:

$$\| \| p - p_h \| \|_{\Omega}^2 = \left\| \mathbb{K}^{\frac{1}{2}} \nabla (p - p_h) \right\|_{\Omega}^2 = \int_{\Omega} (\mathbb{K}^{\frac{1}{2}} \nabla p + \mathbb{K}^{-\frac{1}{2}} \mathbf{t}_h)^2 \quad (7)$$

$$\| \| p - p_h \| \| ^2 \leq \sum_{D \in \mathcal{D}_h} \left( \underbrace{m_D \| f - \operatorname{div} \mathbf{t}_h \|_D}_{\text{residual error}} + \underbrace{\left\| \mathbb{K}^{\frac{1}{2}} \nabla p_h + \mathbb{K}^{-\frac{1}{2}} \mathbf{t}_h \right\|_D}_{\text{flux error}} \right)^2$$

- $m_{D,a} = \frac{C_{P,D} h_D^2}{c_{a,D}}$  if  $D \in D_h^{\text{int}}$ ,  $m_{D,a} = \frac{C_{F,D} h_D^2}{c_{a,D}}$  if  $D \in D_h^{\text{ext}}$
- $C_{P,D}$  is equal  $\frac{1}{\pi^2}$  if  $D$  is convex,  $C_{F,D}$  is equal to 1 on general.

- Effectivity index: 
$$\frac{\left( \sum_{D \in \mathcal{D}_h} (\eta_{R,D} + \eta_{DF,D})^2 \right)^{\frac{1}{2}}}{\| \| p - p_h \| \|_{\Omega}}$$

$T$ : = Triangulation of  $\Omega$ , for all  $\tau \in T$  we define  $v(\tau)$  the "newest vertex".

$E(\tau)$ : = Is the longest edge of  $\tau$ ,  $v(\tau)$  is the vertex opposite to  $E(\tau)$ .

- (R1): The first step consists in dividing the elements by joining  $v(\tau)$  to the middle  $I$  of  $E(\tau)$ .
- (R2):  $I$  becomes the "newest vertex" of each of the two created triangles.
- (R3): Neighbor refinement by  $R1$  and conformity.

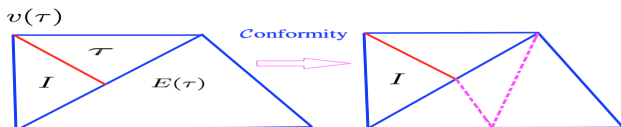
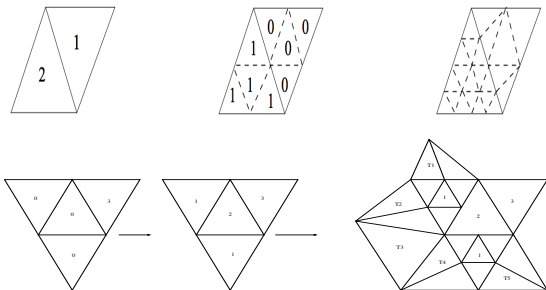


Figure: Bisect a triangle and Completion by Newest-Vertex-Bisection strategy



**Figure:** Mesh refinement with ADAPT and conformity with propagation levels

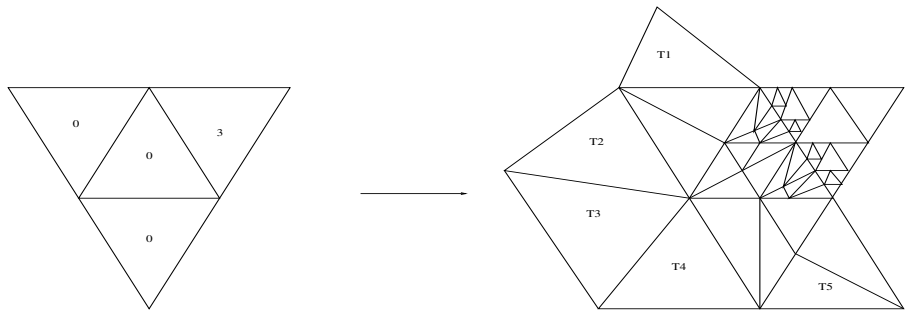


Figure: Mesh refinement with ADAPT-NEWEST

- First we proceed in a first time by refining our mesh by the ADAPT strategy, then for the conformity one uses the method Newest vertex bisection.
- There is no more propagation of the refinement on triangles  $T_1$ ,  $T_2$  et  $T_4$ .

# Test with an analytical solution, $\alpha = 0,127$



- Problem:

$$\begin{aligned} -\operatorname{div}(\mathbb{K}\nabla p) &= f \quad \text{in } \Omega = (-1, 1)^2 \\ p &= 0 \quad \text{on } \partial\Omega \end{aligned}$$

- heterogeneous permeability

$$\mathbb{K} = \begin{cases} 1 \cdot \mathbb{I}_2 & \text{if } x \in \Omega_{1,4} \\ 100 \cdot \mathbb{I}_2 & \text{else.} \end{cases}$$

- Solution

$$p \in H^{1+\alpha}(\Omega), \quad a_i, b_i = \text{const.}$$

$$p(r, \theta) = r^\alpha (a_i \sin(\alpha\theta) + b_i \cos(\alpha\theta))$$

Regular mesh:  $\alpha = 0.127$



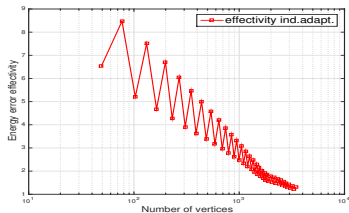
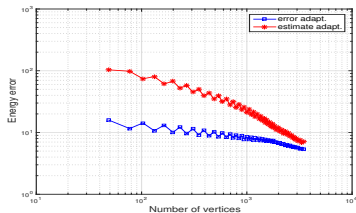


Figure: Energy Error, Estimator, Efficiency

### NewestVB approach

iter	DoFs	$\eta$	$\epsilon_1$	$\epsilon_2$	$f_\eta$	CPU
1	128	103.3915	15.836	0.30586	6.5289	0.679237
6	436	67.4077	10.0475	0.13689	6.7089	0.668070
12	942	44.08	8.8284	0.074364	4.993	0.074364
24	2170	18.4356	7.4593	0.02977	2.4715	1.426176
59	7162	7.0814	5.3987	0.021182	1.3117	4.063058
Total						<b>75.13</b>

Regular mesh :  $\alpha = 0.127$

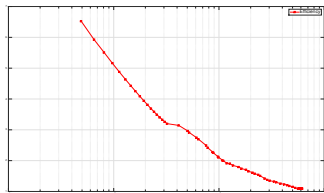
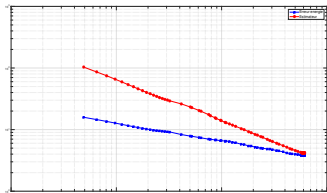


Figure: Estimateur, erreur energie (gauche), efficacité (droite)

#### AdaptNVB approach

iter	DoFs	$\eta$	$\epsilon_1$	$\epsilon_2$	$f_\eta$	CPU
1	240	103.3915	15.836	0.30586	6.5289	0.671513
6	1040	42.7369	10.677	0.081534	4.0027	0.918080
13	2160	20.8044	7.846	0.035467	2.4715	1.462513
24	4920	9.3623	5.8007	0.022808	1.614	2.870125
29	7296	6.9263	5.104	0.022277	1.357	3.995373
Total						<b>35.13</b>

# Irregular mesh : $\alpha = 0.127$

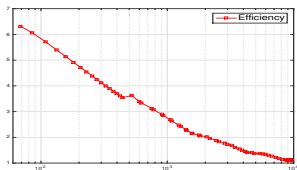
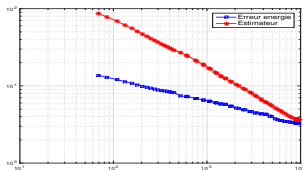


Figure: Estimateur, erreur energie (gauche), efficacité (droite)

# Nested adaptive vertex-centered finite volume

The diffusion in a two-dimensional closed medium  $\Omega \subset \mathbb{R}^2$  with boundary  $\partial\Omega$  is described by the following equation

$$\begin{aligned} -\nabla \cdot (\mathbb{K} \nabla u(\mathbf{x})) &= f(\mathbf{x}), & \forall \mathbf{x} \in \Omega, \\ u(\mathbf{x}) &= g(\mathbf{x}), & \forall \mathbf{x} \in \partial\Omega, \end{aligned} \tag{8}$$

where  $f$  is the external force,  $g$  the boundary source, and  $\mathbb{K}$  is a piecewise constant diffusion coefficient.

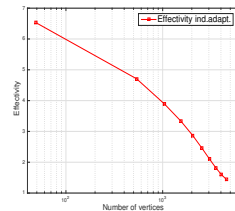
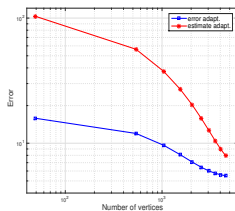
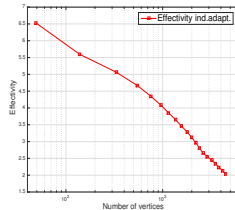
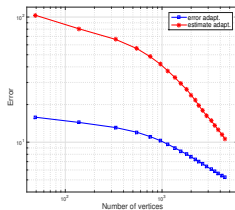


Figure: Results using conventional approach (left) and using nested approach (right).

**Table:** Comparison between the conventional and nested approaches. CPU times are in seconds.

Conventional approach

iter	DoFs	$\eta$	$\epsilon_1$	$\epsilon_2$	$f_\eta$	CPU
1	240	103.3915	15.836	0.30586	6.5289	0.71
6	2256	42.0556	10.298	0.079115	4.0838	1.45
20	9903	10.617	5.2289	0.020684	2.0305	5.5
Total						<b>60.691</b>

Nested approach

iter	DoFs	$\eta$	$\epsilon_1$	$\epsilon_2$	$f_\eta$	CPU
1	1046	103.3915	15.836	0.30586	6.5289	0.39
3	3054	37.5234	9.6343	0.065962	3.8948	1.5
10	10082	7.9307	5.5007	0.024378	1.4418	6.8
Total						<b>35.13</b>

# Algorithm

Parameter	Signification
$\mathcal{I}_{div}$	Table have the values 0 or 1, 1: the triangle that contains this vertice must be refined, and 0 otherwise
$\mathcal{M}_{marker}$	Table that indicates whether an edge should be marked or not.
$\mathcal{N}_{Lev}$	maximum number of multi-level refinement.
$\mathcal{N}_{div}$	contains 0 or 1, 0: the edge is not yet divided, 1: the node is already created in the middle of the edge.
$\mathcal{N}_{Ref}$	Maximum level of refinement.

- First step construction of dual mesh  $\mathcal{D}_h$ , simplicial mesh  $\mathcal{S}_h$ .
- We compute the estimators on the simplicial mesh edges which will allow us to calculate the estimator in the node that surrounds these edges.
- With the aid of this estimator, a threshold is defined in order to obtain the  $i_{div}$  in the node.
- After calculating  $i_{div}$  in the cell (this information will allow us to know if the cell must be adapted or not and also know its level of refinement.



- If for example the level of refinement is 3 one proceeds as follows:  
We will iterate 3 times and in each iteration we will make an adaptation (Adapt strategy) and the conformity (new-vertex).

Parameter	Signification
$\mathcal{I}_{div}$	Table have the values 0 or 1, 1: the triangle that contains this vertice must be refined, and 0 otherwise
$\mathcal{M}_{marker}$	Table that indicates whether an edge should be marked or not.
$\mathcal{N}_{Lev}$	maximum number of multi-level refinement.
$\mathcal{N}_{div}$	contains 0 or 1, 0: the edge is not yet divided, 1: the node is already created in the middle of the edge.
$\mathcal{N}_{Ref}$	Maximum level of refinement.

---

---

```
1 for  $t:=1$  to  $\mathcal{N}_{Ref}$  do
2   Creation of all data structures
3   Calcul of numerical solution
4   calcul of estimator
5   Calcul of exact error
6   Compute of criteria
7   for  $k:=1$  to  $\mathcal{N}_{Lev}$  do
8     Mesh refinement with Adapt strategy
9     Conformity with NewestVB startegy
10    Marking of new triangles created to be refined
11  end
12 end
```

---

# Conclusion

- Coupled method Adpat-Newest to refine the mesh.
- Convergence of error to .
- Performance of CPU time using multi-level adaptation.

**THANK YOU FOR YOUR ATTENTION !!**