



UNIVERSITÀ
DEGLI STUDI
FIRENZE

DIMAI

DIPARTIMENTO DI MATEMATICA
E INFORMATICA "ULISSE DINI"



Funded by the
European Union
NextGenerationEU



Italiadomani

PIANO NAZIONALE DI RIPRESA E RESILIENZA

CN1 - Spoke 6



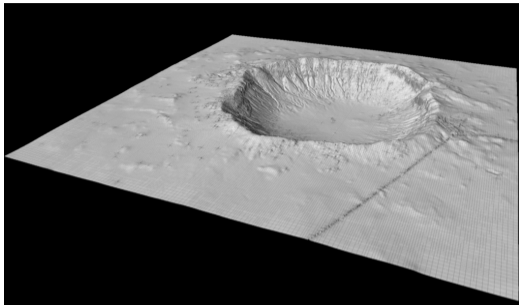
An Algorithmic Introduction to LR B-splines

Francesco Patrizi

Picture: Acropolis

Athens, Greece

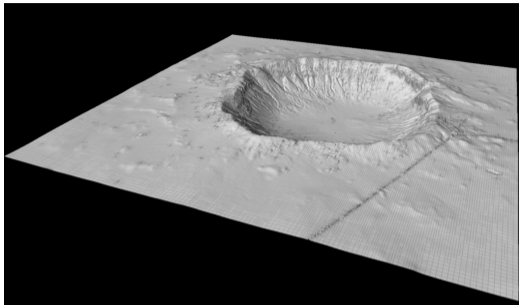
A tool for adaptive approximations



Quasi-Interpolation, Meteor Crater, AZ

Wind streamlines around a telescope

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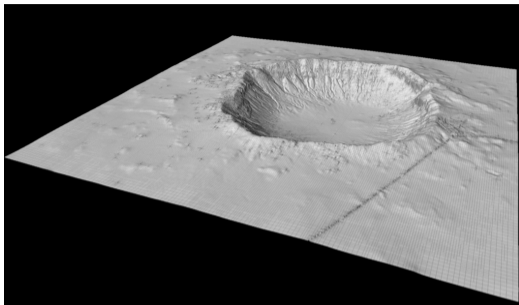


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T. Dokken



T. Lyche



K. F. Pettersen

Recalling univariate B-splines

(Local) Knot Vector: Given a degree p , $\mathbf{t} = \mathbf{t}_p$ with $|\mathbf{t}_p| = p + 2$ with repetitions

$$\underbrace{t_1 = \cdots = t_{m_1}}_{\max p+1 \text{ times}} < \underbrace{t_{m_1+1} = \cdots = t_{m_1+m_2}}_{\max p+1 \text{ times}} < \cdots$$

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$$B[\mathbf{t}](t) = \frac{t - t_1}{t_{p+1} - t_1} B[t_1, \dots, t_{p+1}](t) + \frac{t_{p+2} - t}{t_{p+2} - t_2} B[t_2, \dots, t_{p+2}](t),$$

where each time a fraction with zero denominator appears, it is taken as zero. The initial B-splines of degree 0 are defined as

$$B[t_i, t_{i+1}](t) := \begin{cases} 1 & \text{if } t_i \leq t < t_{i+1}; \\ 0 & \text{otherwise;} \end{cases}$$

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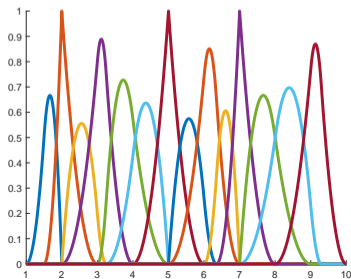
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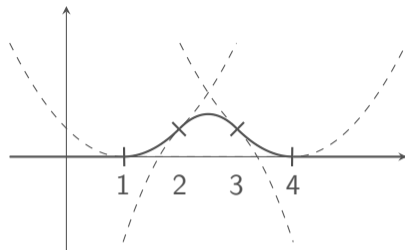
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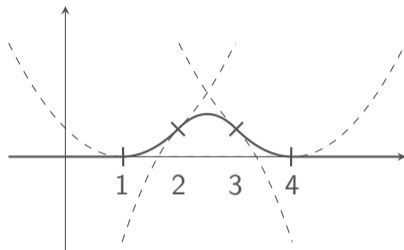


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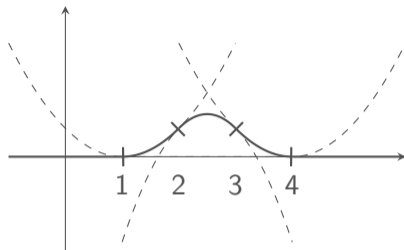
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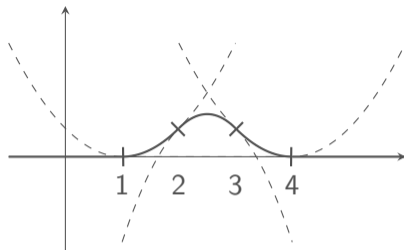
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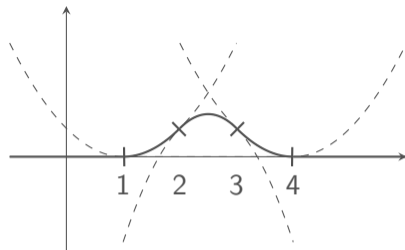
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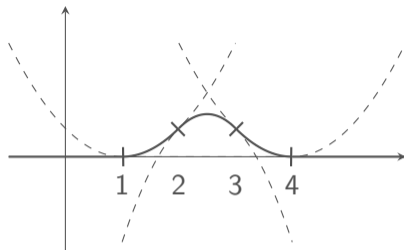
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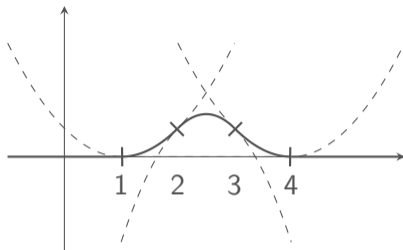
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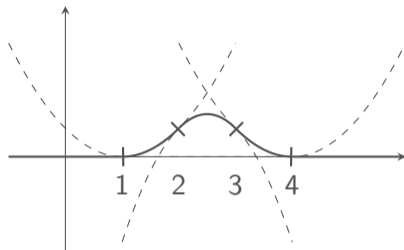
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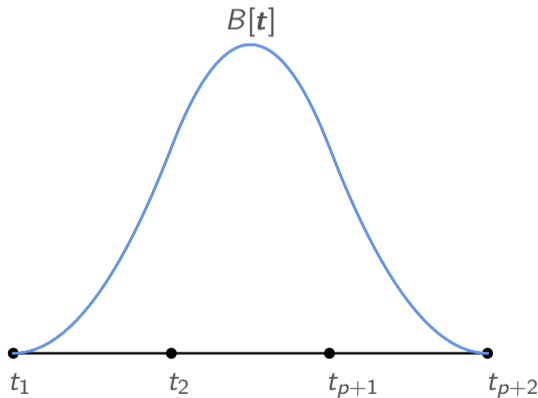
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Knot Insertion

Knot Insertion: Suppose we insert a knot $\hat{t} \rightarrow \mathbf{t}$. We obtain two knot vectors \mathbf{t}_1 and \mathbf{t}_2 , considering the first and the last $p+2$ knots respectively in $(t_1, \dots, \hat{t}, \dots, t_{p+2})$. Then

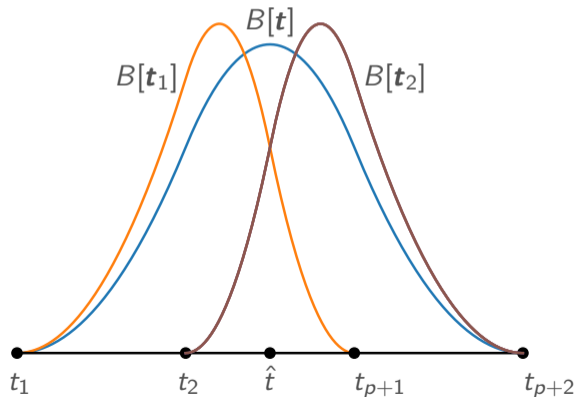
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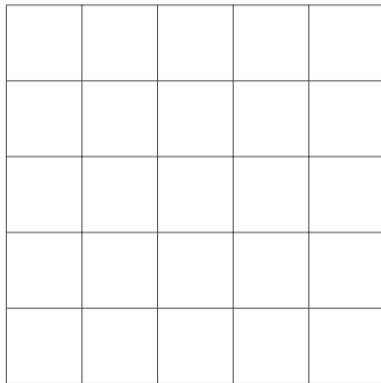
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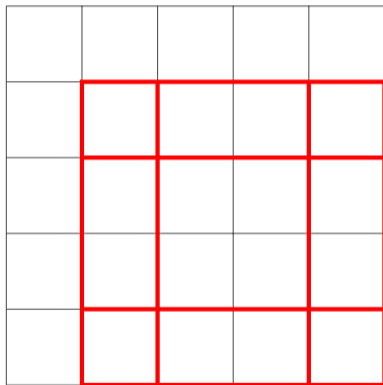
B-splines on Tensor Meshes

Given a tensor mesh \mathcal{N} and a bidegree (p_1, p_2) (for instance $(2, 2)$),



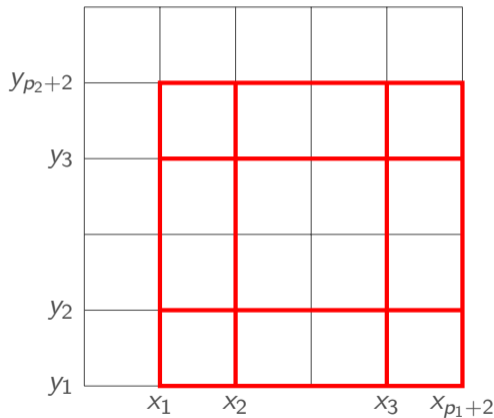
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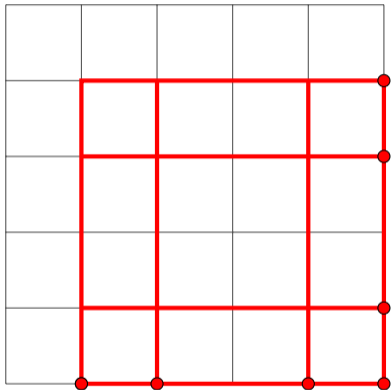
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Such vertical and horizontal lines can be parametrized as $\{x_i\} \times [y_1, y_{p_2+2}]$ and $[x_1, x_{p_1+2}] \times \{y_j\}$ with $\mathbf{x} := (x_i)_{i=1}^{p_1+2}$ and $\mathbf{y} = (y_j)_{j=1}^{p_2+2}$.

B-splines on Tensor Meshes

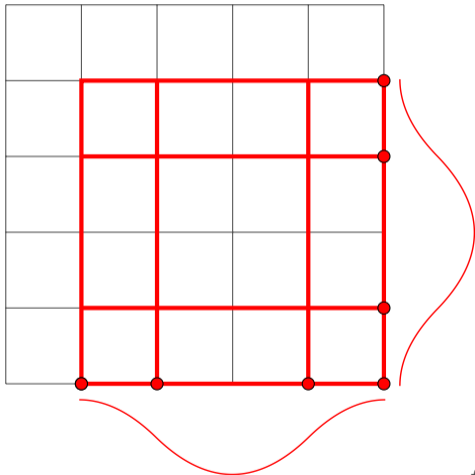
x and y are knot vectors on top on which we define univariate B-splines of degrees p_1 and p_2 .



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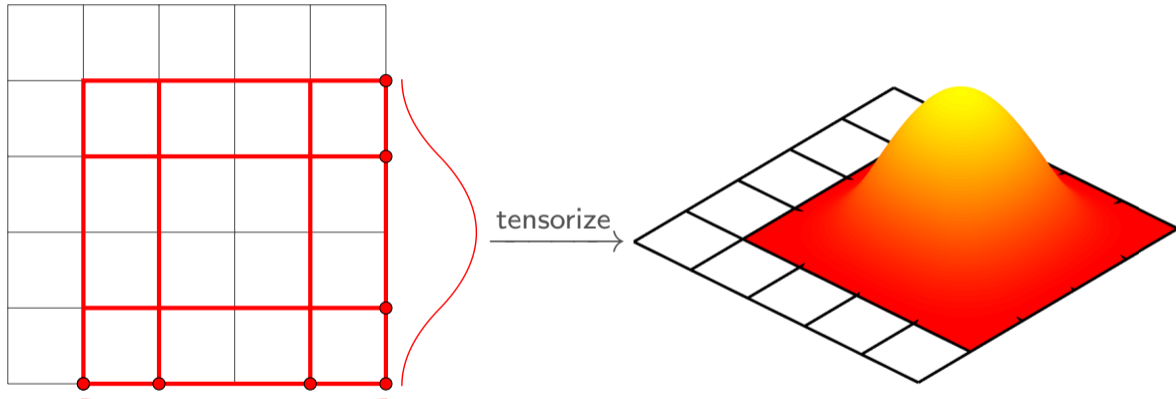
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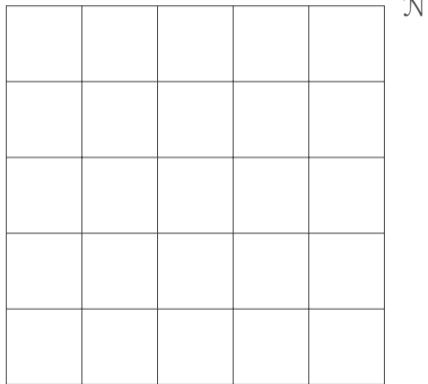


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Each \mathcal{N}_B corresponds to a B-spline B defined on \mathcal{N} .

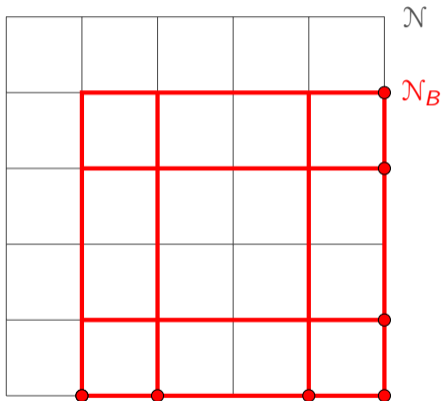
If no line in $\mathcal{N} \setminus \mathcal{N}_B$ traverses $\text{int}(\text{supp } B)$ then we say that B has minimal support on \mathcal{N} .



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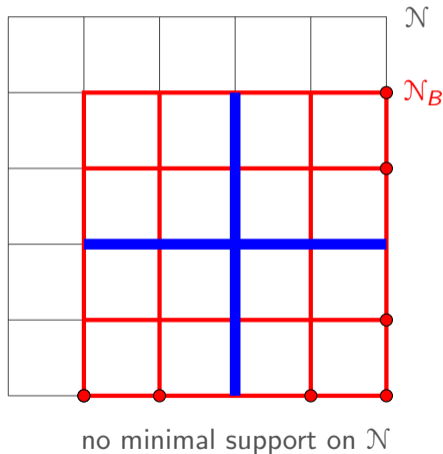
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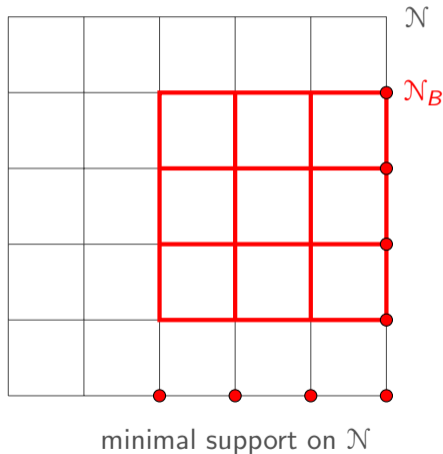
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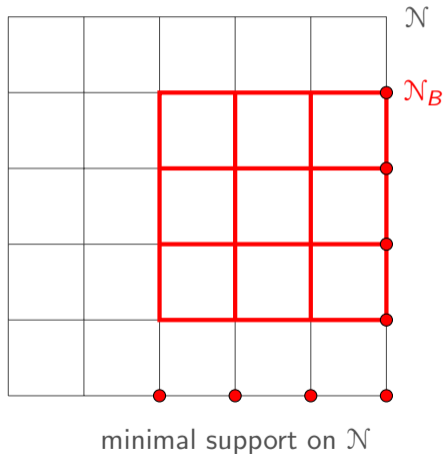
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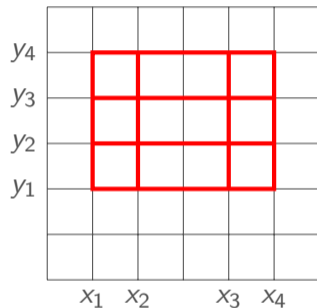


We call B-spline set on \mathcal{N} the set of all the minimal support B-splines on \mathcal{N} .

Knot insertion

Assume $B[\mathbf{x}, \mathbf{y}]$ no minimal support on \mathcal{N} because of a vertical line at $x = \hat{x}$ with $\hat{x} \notin \mathbf{x}$.

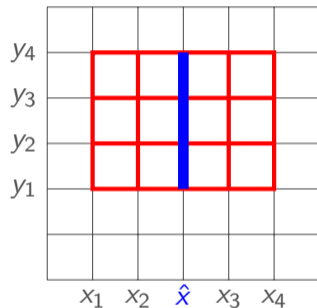
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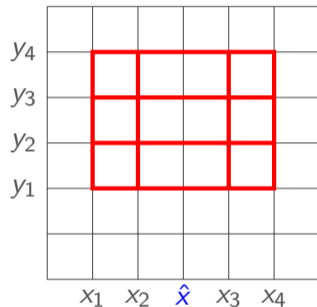
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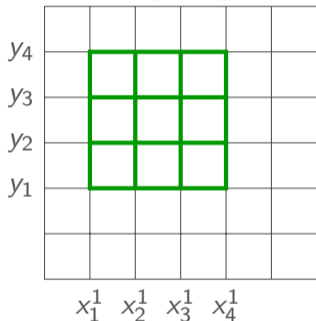
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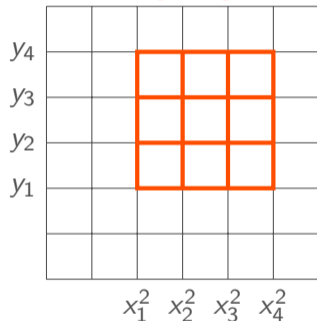
$= \alpha^1 \cdot$

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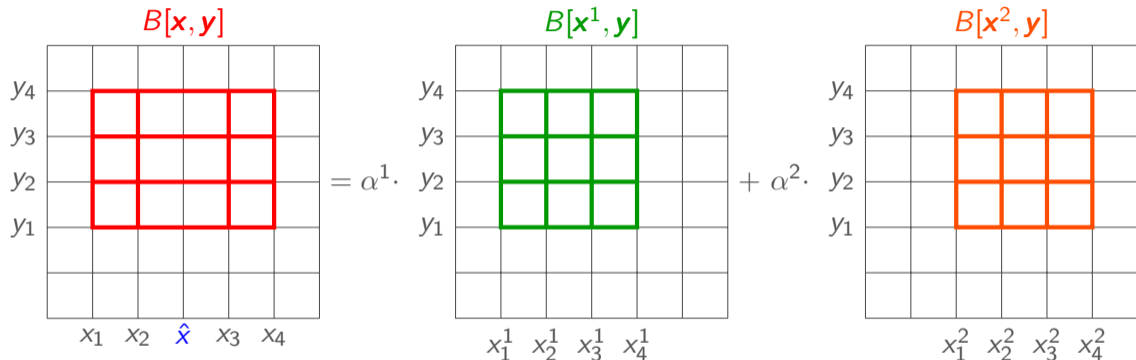
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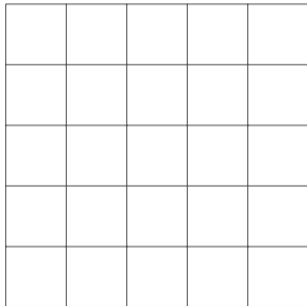
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with $\alpha^1, \alpha^2 \in (0, 1]$. $B[\mathbf{x}, \mathbf{y}]$ is expressed in terms of B-splines of minimal support on \mathcal{N} .

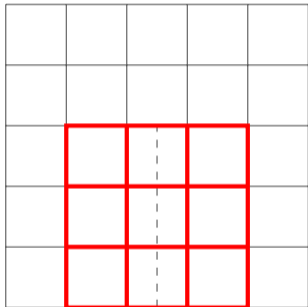
LR meshes and LR B-splines

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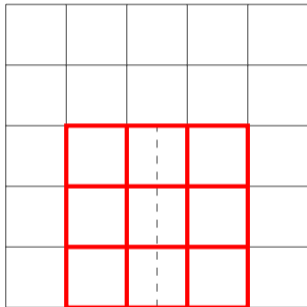
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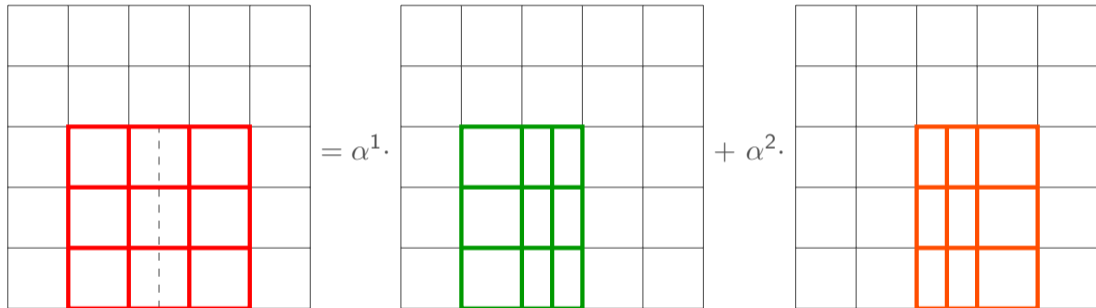
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By construction B has not minimal support on the new mesh $\mathcal{N}' = \mathcal{N} \cup \gamma$. By knot insertion we replace B with the B-splines B^1 and B^2 of minimal support on \mathcal{N}' . This operation creates a new set \mathcal{B}' of minimal support B-splines on \mathcal{N}' .

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LR mesh \mathcal{N}' (recursive definition):

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- ▶ a tensor mesh,
- ▶ obtained by insertion of a new line from the LR mesh \mathcal{N} , traversing at least one support.

LR B-spline set \mathcal{B}' (recursive definition): It is either

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with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.

LR meshes and LR B-splines

Partition of Unity Weights: sum of the knot insertion coefficients

LR meshes and LR B-splines

Partition of Unity Weights: sum of the knot insertion coefficients

$$\begin{aligned}
 & \begin{array}{|c|c|c|c|c|} \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline \end{array} = \alpha^1 \cdot \begin{array}{|c|c|c|c|c|} \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline \end{array} + \alpha^2 \cdot \begin{array}{|c|c|c|c|c|} \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline \end{array} \\
 & \begin{array}{|c|c|c|c|c|} \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline \end{array} = \beta^1 \cdot \begin{array}{|c|c|c|c|c|} \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline \end{array} + \beta^2 \cdot \begin{array}{|c|c|c|c|c|} \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline \end{array}
 \end{aligned}$$

The PoU weight for

$$B = \begin{array}{|c|c|c|c|c|} \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline & & & & \\ \hline \end{array}$$

is $w_B = \alpha^1 + \beta^1$.

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Partition of Unity Weights: sum of the knot insertion coefficients

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 & \begin{array}{|c|c|c|c|} \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline \end{array} = \beta^1 \cdot \begin{array}{|c|c|c|c|} \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline \end{array} + \beta^2 \cdot \begin{array}{|c|c|c|c|} \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline \end{array}
 \end{aligned}$$

The PoU weight for

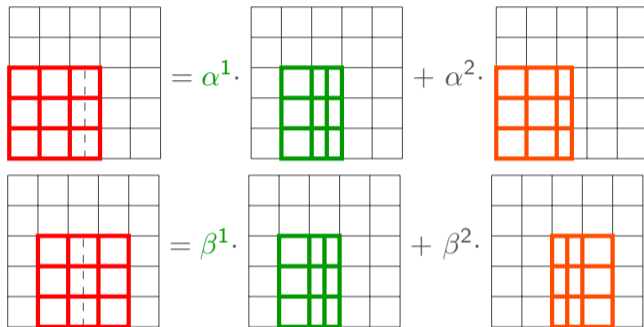
$$B = \begin{array}{|c|c|c|c|} \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline & & & \\ \hline \end{array}$$

is $w_B = \alpha^1 + \beta^1$.

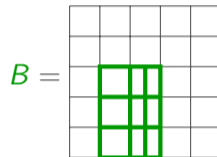
Remark: If not specified otherwise, we consider internal meshlines of multiplicity 1 and boundary meshlines of multiplicity $p_k + 1$, for $k = 1, 2$, for vertical and horizontal meshlines respectively.

LR meshes and LR B-splines

Partition of Unity Weights: sum of the knot insertion coefficients



The PoU weight for



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Remark: Meshline insertion ordering can often be changed. However, the final LR B-spline set is well defined because independent of such insertion ordering.

LR meshes and LR B-splines

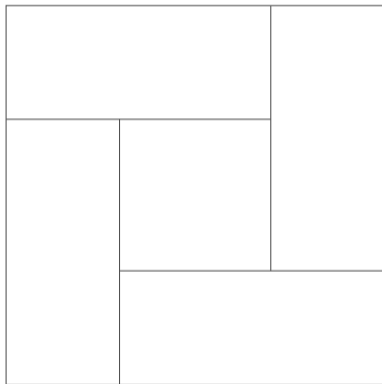
Remark: Not all meshes with local lines are LR meshes

LR mesh: $\begin{cases} \mathcal{N}_{i+1} = \mathcal{N}_i \cup \gamma_i & \gamma_i \text{ new line, traversing at least one support} \\ \mathcal{N}_0 & \text{tensor mesh,} \end{cases}$

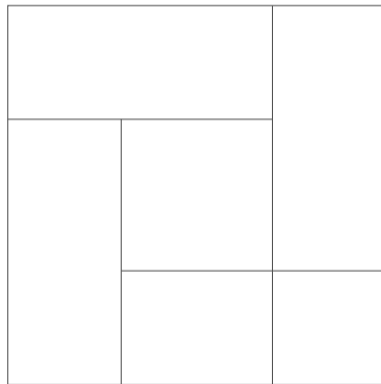
LR meshes and LR B-splines

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Not LR mesh



LR mesh

LR meshes and LR B-splines

Remark: Secondary splits may be needed.

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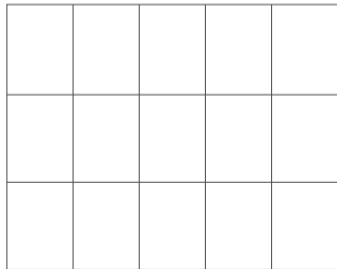
LR meshes and LR B-splines

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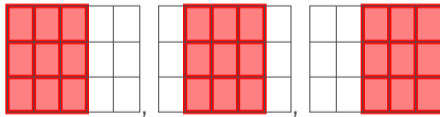
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(a) current mesh



(b) \mathcal{B}_0

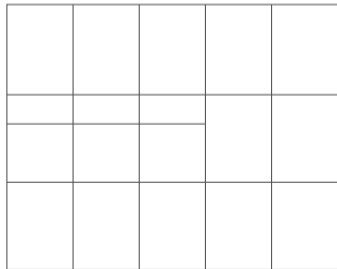
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

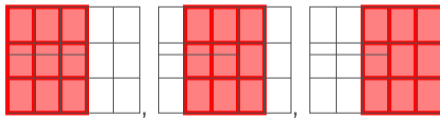
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(a) current mesh



(b) \mathcal{B}_0

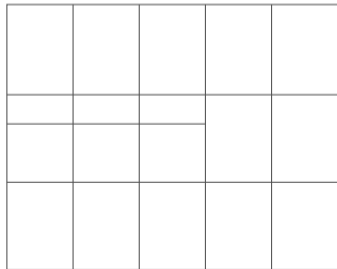
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

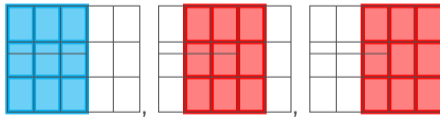
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(a) current mesh



(b) \mathcal{B}_0

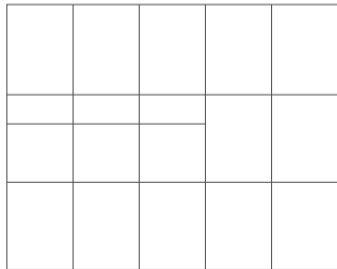
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

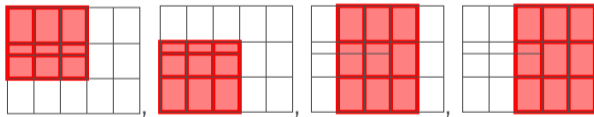
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(a) current mesh



(b) \mathcal{B}_1

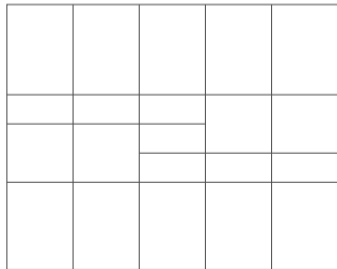
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

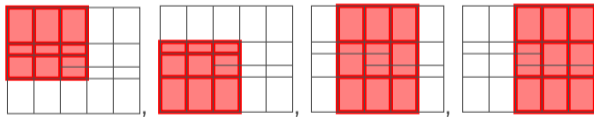
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with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_1

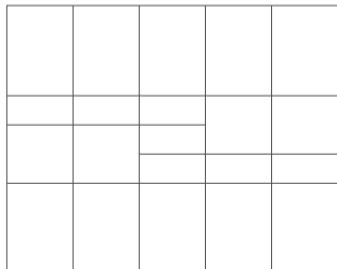
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

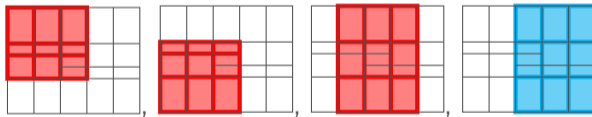
LR mesh: $\begin{cases} \mathcal{N}_{i+1} = \mathcal{N}_i \cup \gamma_i & \gamma_i \text{ new line, traversing at least one support} \\ \mathcal{N}_0 & \text{tensor mesh,} \end{cases}$

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with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_1

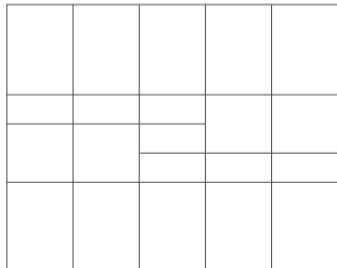
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

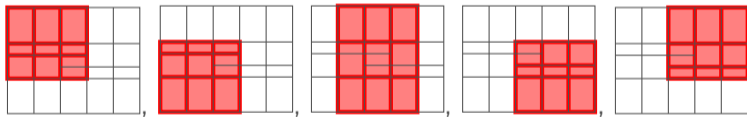
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with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_2

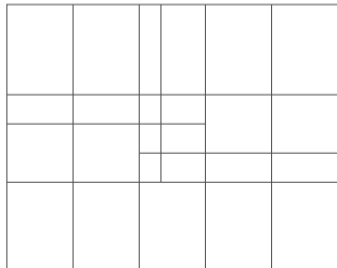
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

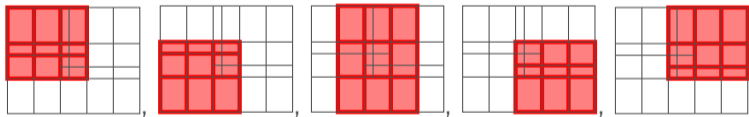
LR mesh: $\begin{cases} \mathcal{N}_{i+1} = \mathcal{N}_i \cup \gamma_i & \gamma_i \text{ new line, traversing at least one support} \\ \mathcal{N}_0 & \text{tensor mesh,} \end{cases}$

LR B-splines: $\begin{cases} \mathcal{B}_{i+1} = (\mathcal{B}_i \setminus \mathcal{B}_i(\gamma_i)) + \mathcal{K}(\mathcal{B}_i(\gamma_i)) & \mathcal{B}_i(\gamma_i) := \text{B-splines in } \mathcal{B}_i \text{ traversed by } \gamma_i \\ \mathcal{B}_0 & \text{tensor B-splines on } \mathcal{N}_0 \end{cases}$

with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_2

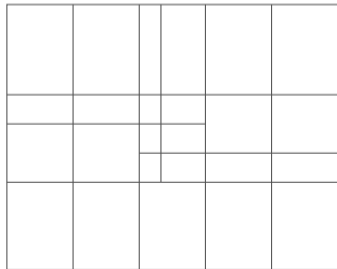
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

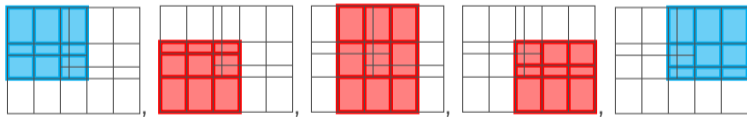
LR mesh:
$$\begin{cases} \mathcal{N}_{i+1} = \mathcal{N}_i \cup \gamma_i & \gamma_i \text{ new line, traversing at least one support} \\ \mathcal{N}_0 & \text{tensor mesh,} \end{cases}$$

LR B-splines:
$$\begin{cases} \mathcal{B}_{i+1} = (\mathcal{B}_i \setminus \mathcal{B}_i(\gamma_i)) + \mathcal{K}(\mathcal{B}_i(\gamma_i)) & \mathcal{B}_i(\gamma_i) := \text{B-splines in } \mathcal{B}_i \text{ traversed by } \gamma_i \\ \mathcal{B}_0 & \text{tensor B-splines on } \mathcal{N}_0 \end{cases}$$

with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_2

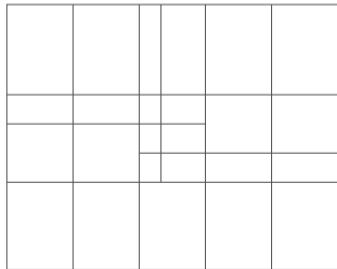
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

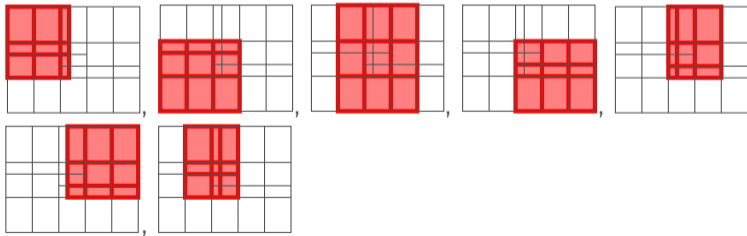
LR mesh: $\begin{cases} \mathcal{N}_{i+1} = \mathcal{N}_i \cup \gamma_i & \gamma_i \text{ new line, traversing at least one support} \\ \mathcal{N}_0 & \text{tensor mesh,} \end{cases}$

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with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_3

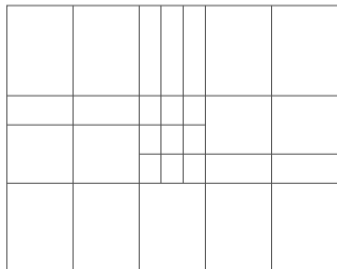
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

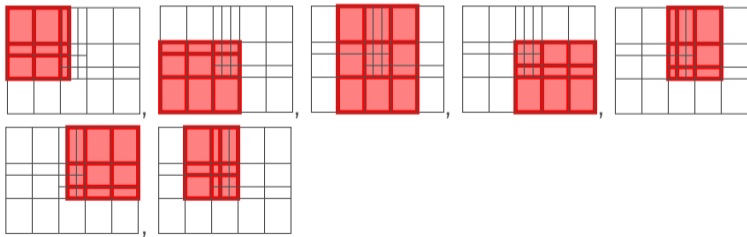
LR mesh: $\begin{cases} \mathcal{N}_{i+1} = \mathcal{N}_i \cup \gamma_i & \gamma_i \text{ new line, traversing at least one support} \\ \mathcal{N}_0 & \text{tensor mesh,} \end{cases}$

LR B-splines: $\begin{cases} \mathcal{B}_{i+1} = (\mathcal{B}_i \setminus \mathcal{B}_i(\gamma_i)) + \mathcal{K}(\mathcal{B}_i(\gamma_i)) & \mathcal{B}_i(\gamma_i) := \text{B-splines in } \mathcal{B}_i \text{ traversed by } \gamma_i \\ \mathcal{B}_0 & \text{tensor B-splines on } \mathcal{N}_0 \end{cases}$

with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_3

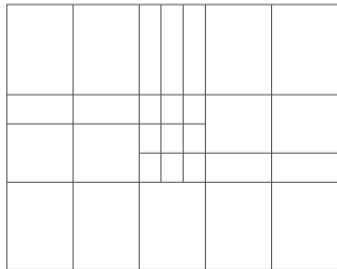
LR meshes and LR B-splines

Remark: Secondary splits may be needed.

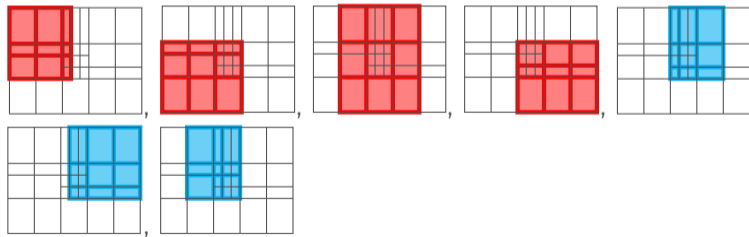
LR mesh: $\begin{cases} \mathcal{N}_{i+1} = \mathcal{N}_i \cup \gamma_i & \gamma_i \text{ new line, traversing at least one support} \\ \mathcal{N}_0 & \text{tensor mesh,} \end{cases}$

LR B-splines: $\begin{cases} \mathcal{B}_{i+1} = (\mathcal{B}_i \setminus \mathcal{B}_i(\gamma_i)) + \mathcal{K}(\mathcal{B}_i(\gamma_i)) & \mathcal{B}_i(\gamma_i) := \text{B-splines in } \mathcal{B}_i \text{ traversed by } \gamma_i \\ \mathcal{B}_0 & \text{tensor B-splines on } \mathcal{N}_0 \end{cases}$

with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_3

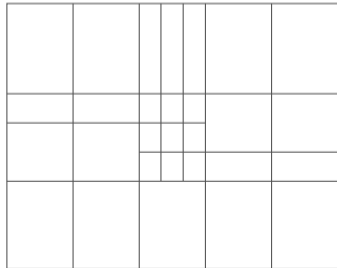
LR meshes and LR B-splines

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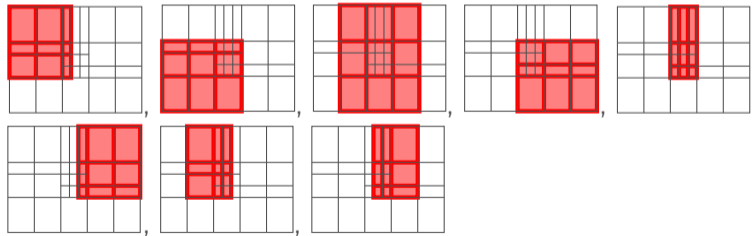
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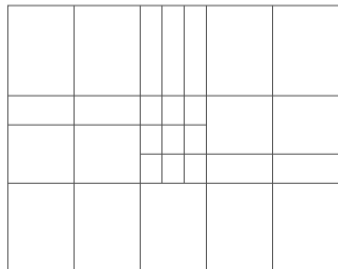
LR meshes and LR B-splines

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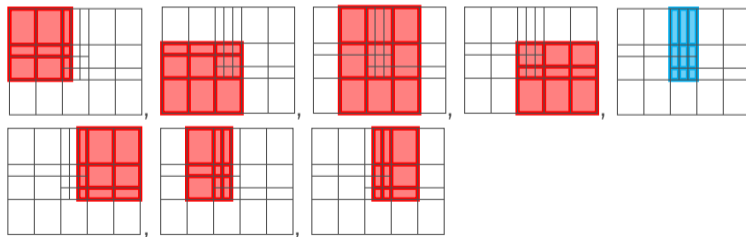
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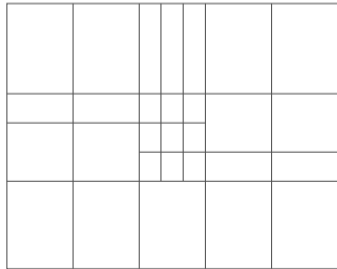
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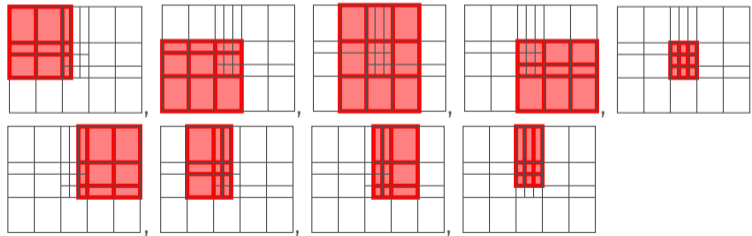
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with \mathcal{K} all the refinements via knot insertion needed to have minimal supports.



(a) current mesh



(b) \mathcal{B}_4

LR meshes and LR B-splines

Remark: LR B-spline set \neq Minimal Support B-spline set.

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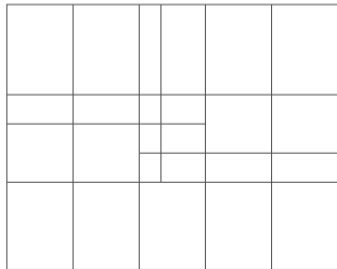
LR meshes and LR B-splines

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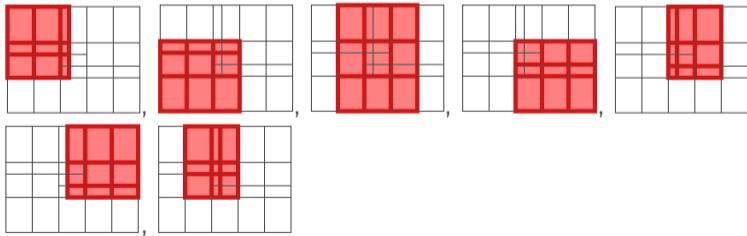
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(a) current mesh



(b) \mathcal{B}_3

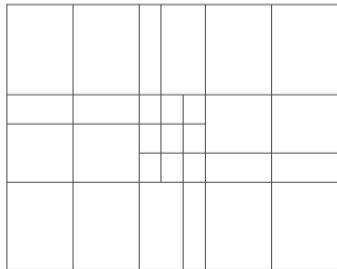
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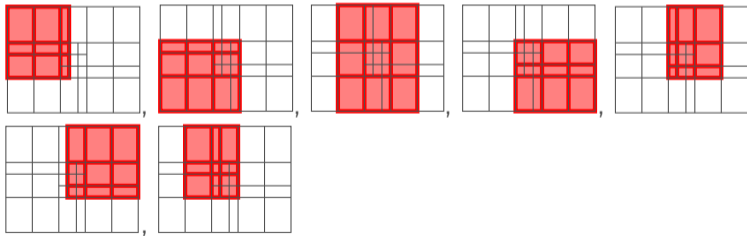
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(a) current mesh



(b) \mathcal{B}_3

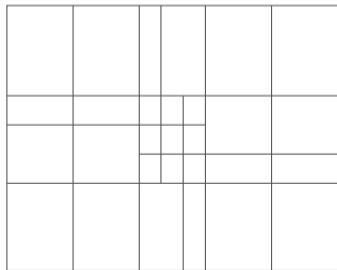
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Remark: LR B-spline set \neq Minimal Support B-spline set.

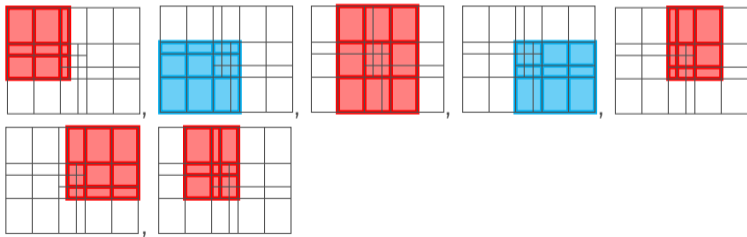
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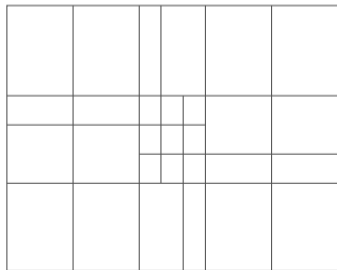
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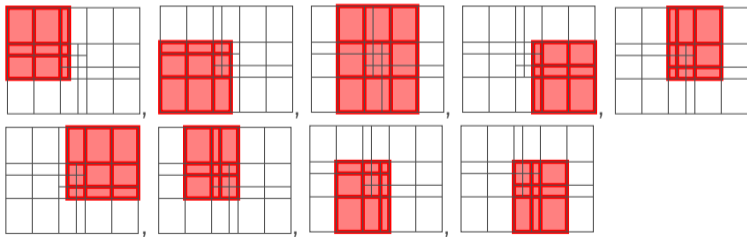
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(a) current mesh



(b) \mathcal{B}_4

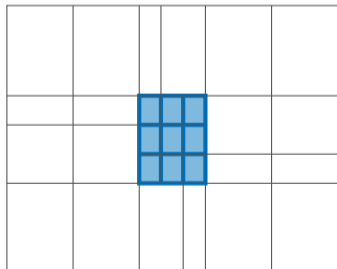
LR meshes and LR B-splines

Remark: LR B-spline set \subsetneq Minimal Support B-spline set.

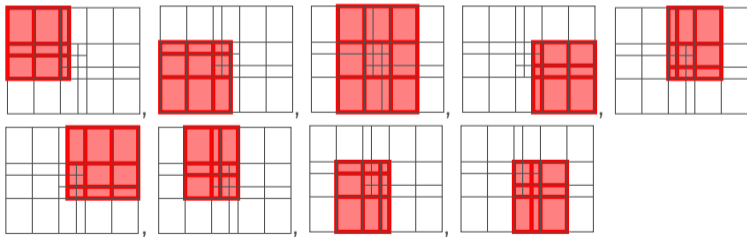
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(a) current mesh



(b) \mathcal{B}_4

Minimum Span Refinement strategy

Input: Bunch of boxes where a larger error is committed in some sense.
For each of such boxes

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1. Among all the LR B-splines on that box, select those with smallest support (semi-perimeter),

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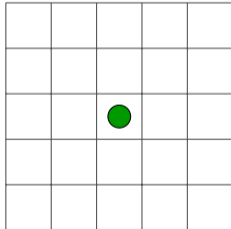
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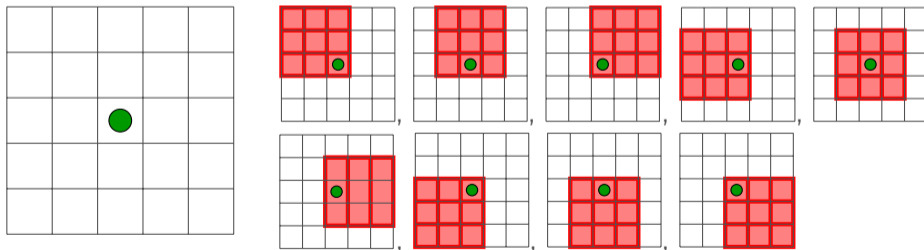


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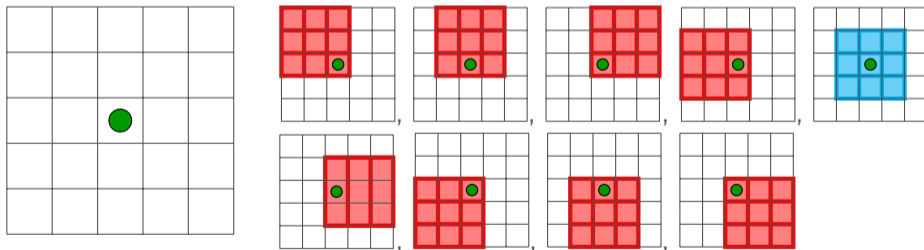


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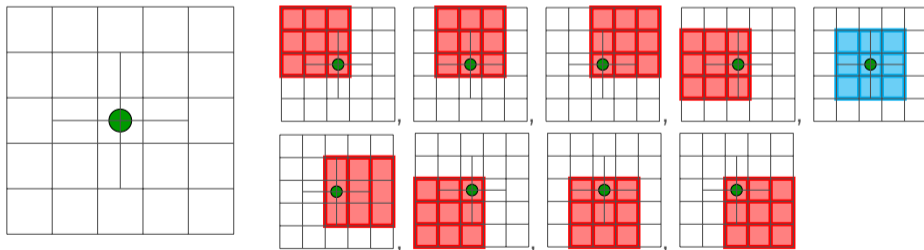


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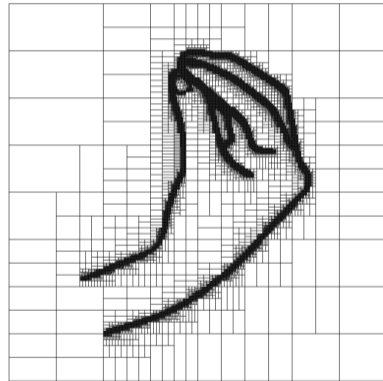
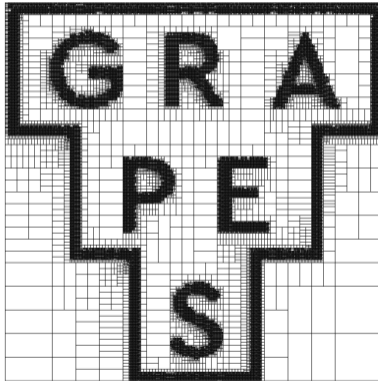
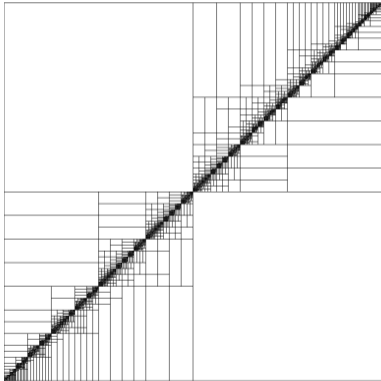
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Minimum Span Refinement strategy: Examples



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For each of such boxes

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Input: Bunch of boxes where a larger error is committed in some sense.

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1. Select all the LR B-splines on that box,
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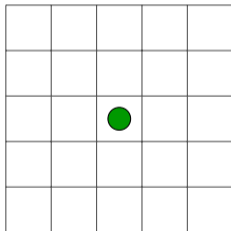
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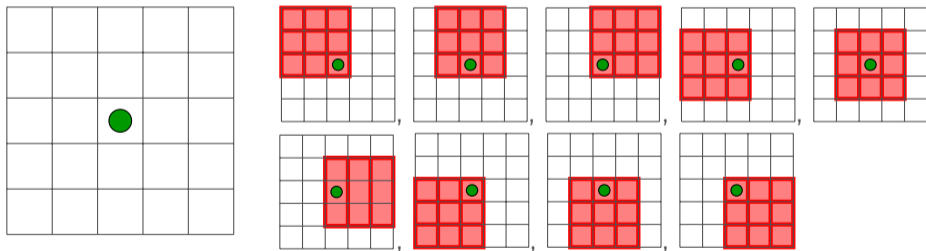
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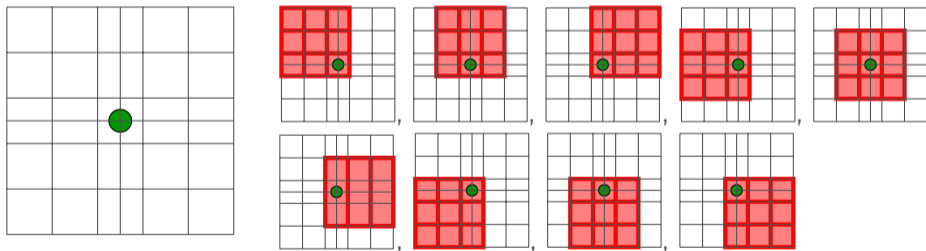
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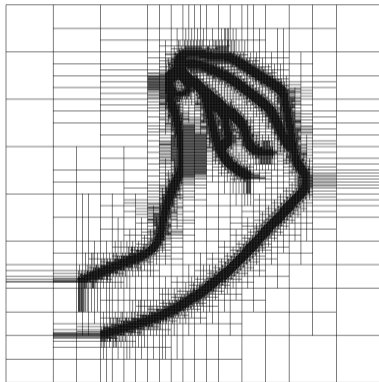
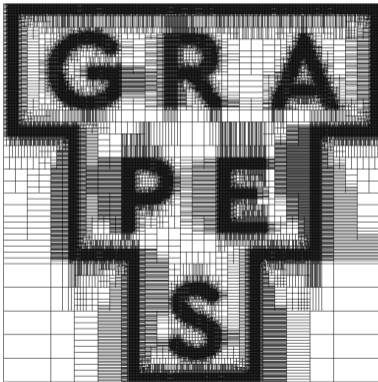
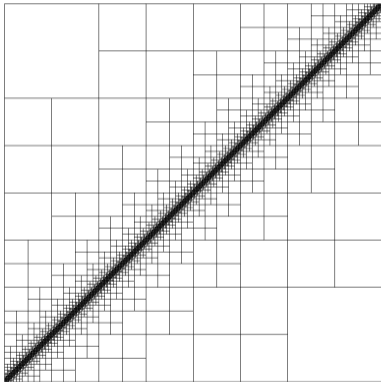
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Full Span Refinement strategy: Examples



Structured Mesh Refinement strategy

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1. Split in 4 all the boxes in their tensor mesh

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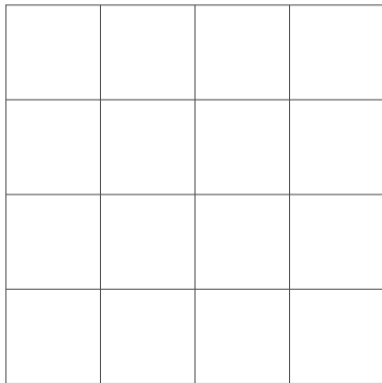
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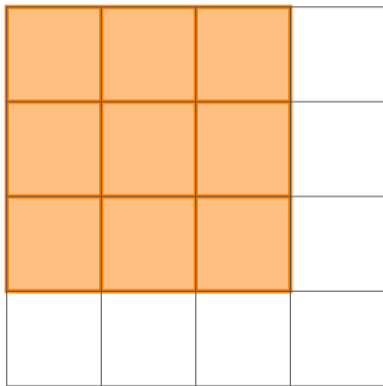
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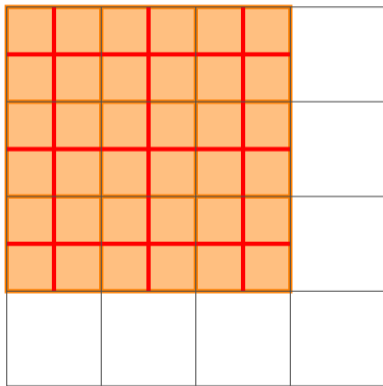
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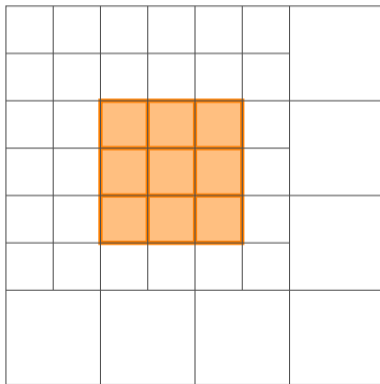
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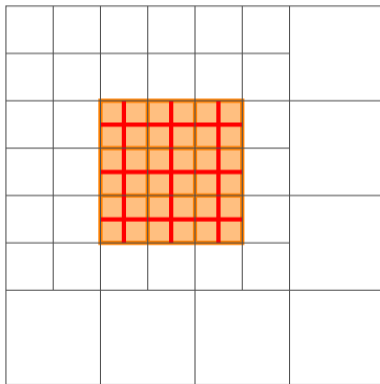
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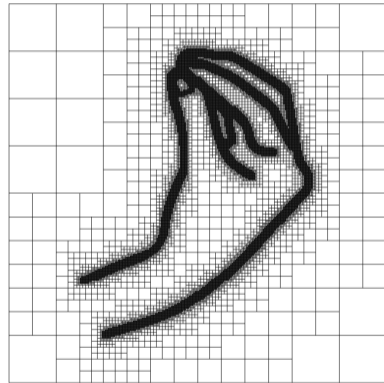
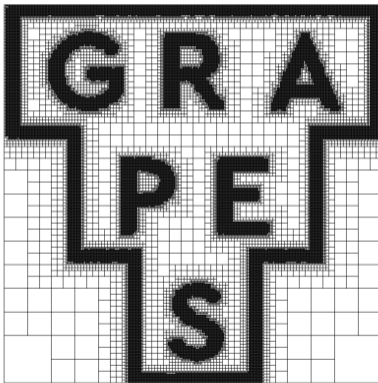
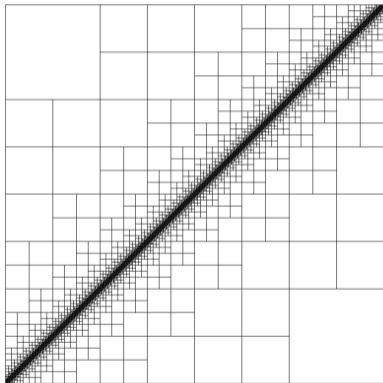
Structured Mesh Refinement strategy

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Structured Mesh Refinement strategy: Examples

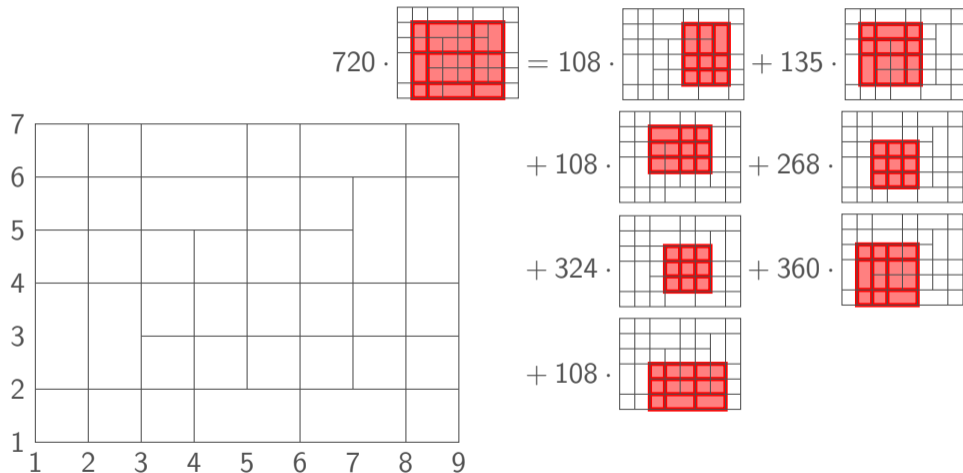


The Linear Dependence Problem

Unfortunately, linear dependence relations may arise in the LR B-spline set.

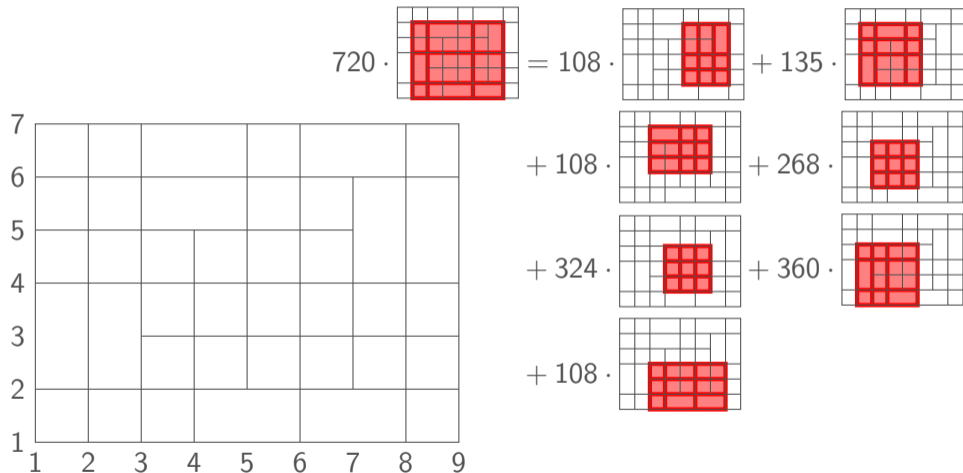
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Minimum Span, Full Span and Structured Mesh may have linear dependence

Conjecture: the latter only for $(p_1, p_2) \geq (4, 4)$.

Seek and Destroy Linear Dependence: Peeling Algorithm

Remark: In every box we span the polynomial space $\Pi_{\mathbf{p}} \Rightarrow$ each box is in at least $(p_1 + 1)(p_2 + 1) = \dim \Pi_{\mathbf{p}}$ LR B-spline supports.

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Remark: In every box we span the polynomial space $\Pi_{\mathbf{p}} \Rightarrow$ each box is in at least $(p_1 + 1)(p_2 + 1) = \dim \Pi_{\mathbf{p}}$ LR B-spline supports.

Overloaded box: box contained in more than $(p_1 + 1)(p_2 + 1)$ LR B-spline supports.

g	g	g	g	g	g	g	g
g	g	g	g	g	g	g	g
	g	10	10 10 10 10 9	g	g	g	g
g	g	10	10 11 12 12 10 10 10 10 10 10 10 12 12 11 10	10	g	g	g
	g	10	10 11 11 11 10 10 9	g	g	g	g
g	g	10	10 11 11 10 10 9	g	g	g	g
	g	10	10 11 11 10 10 9	g	g	g	g
g	g	10	10 9 9 9 10 10 9	g	g	g	g
	g	10	10 9 9 9 9 9 9	g	g	g	g
g	g	10	10 9 9 9 9 9 9	g	g	g	g
	g	10	10 9 9 9 9 9 9	g	g	g	g
g	g	10	10 9 9 9 9 9 9	g	g	g	g
	g	10	10 9 9 9 9 9 9	g	g	g	g
g	g	10	10 12 11 10 10 11 10 9	g	g	g	g
	g	10	10 12 11 10 10 10 9	g	g	g	g
g	g	10	10 11 11 11 10 10 9	g	g	g	g
	g	10	10 10 10 11 11 11 9	g	g	g	g
g	g	10	10 11 12 12 10 10 10 10 10 10 10 12 12 11 10	10	g	g	g
	g	10	10 10 10 10 9	g	g	g	g
g	g	g	g	g	g	g	g

Seek and Destroy Linear Dependence: Peeling Algorithm

Remark: In every box we span the polynomial space $\Pi_{\mathbf{p}} \Rightarrow$ each box is in at least $(p_1 + 1)(p_2 + 1) = \dim \Pi_{\mathbf{p}}$ LR B-spline supports.

Overloaded box: box contained in more than $(p_1 + 1)(p_2 + 1)$ LR B-spline supports.

g	g	g	g	g	g	g	g
g	g	g	g	g	g	g	g
	g	10	10 10 10 10 9	g	g	g	g
g	g	10	10 11 12 12 10 10 10 10 10 10 10 12 12 11 10	10	g	g	g
	g	10	10 10 10 11 11 11 9	g	g	g	g
g	g	10	10 11 11 10 10 9	g	g	g	g
	g	10	10 12 11 10 10 10 9	g	g	g	g
g	g	10	10 9 9 9 10 10 9	g	g	g	g
	g	10	10 9 9 9 9 9 9	g	g	g	g
g	g	10	10 9 9 9 9 9 9	g	g	g	g
	g	10	10 9 9 9 9 9 9	g	g	g	g
g	g	10	10 9 9 9 9 9 9	g	g	g	g
	g	10	10 9 9 9 9 9 9	g	g	g	g
g	g	10	10 12 11 10 10 11 10 9	g	g	g	g
	g	10	10 12 11 10 10 10 9 9	g	g	g	g
g	g	10	10 11 11 11 10 10 9	g	g	g	g
	g	10	10 10 10 11 11 11 9	g	g	g	g
g	g	10	10 11 12 12 10 10 10 10 10 10 10 12 12 11 10	10	g	g	g
	g	10	10 10 10 10 9 9 9 9 9 9 9 10 10 10 10	10	g	g	g
g	g	g	g	g	g	g	g
	g	g	g	g	g	g	g
g	g	g	g	g	g	g	g

Overloaded LR B-spline: all the boxes in its support are overloaded.

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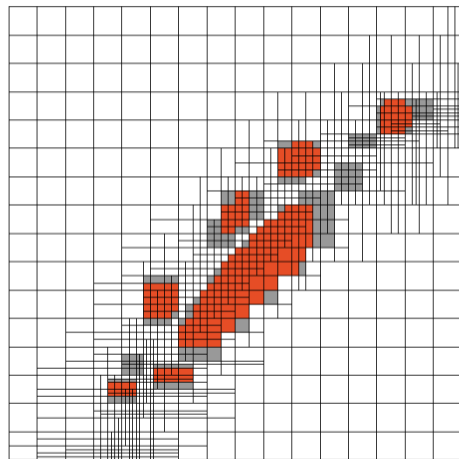
💡 A linear dependence relation needs at least 2 functions.

⇒ If a box is just in one overloaded LR B-spline B , then B cannot be part of a linear dependence relation.

Seek and Destroy Linear Dependence: Peeling Algorithm

Peeling Algorithm

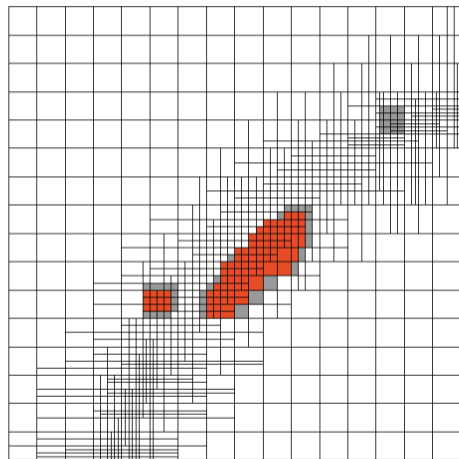
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Peeling Algorithm

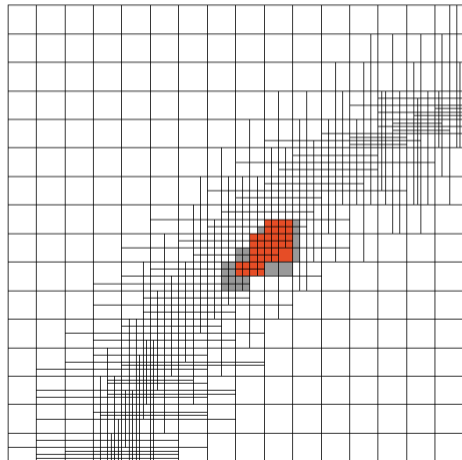
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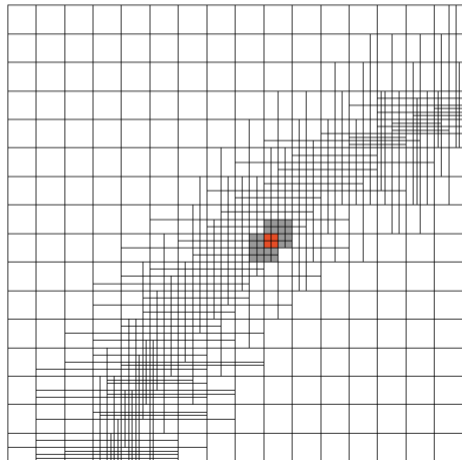
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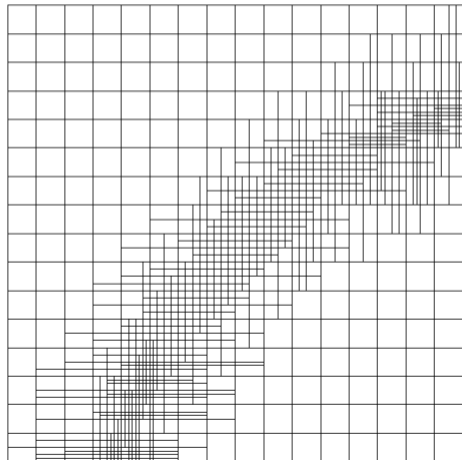
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Conjecture: Peeling Algorithm++ sorts out all cases, if $\mathcal{B}_1^O = \emptyset$ then there is a linear dependence relation.

Local linear independence and N_2S property

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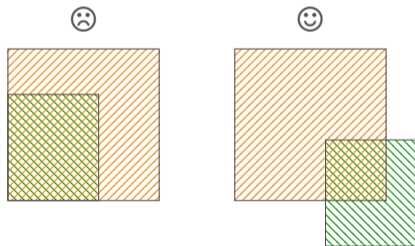
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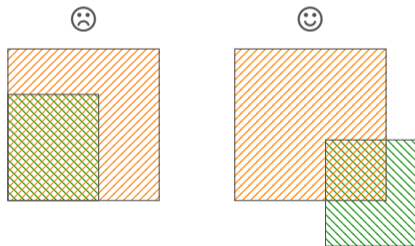
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How do we build LR meshes with the N_2S property?

Non-Nested Support Structured (N_2S_2) Mesh Refinement



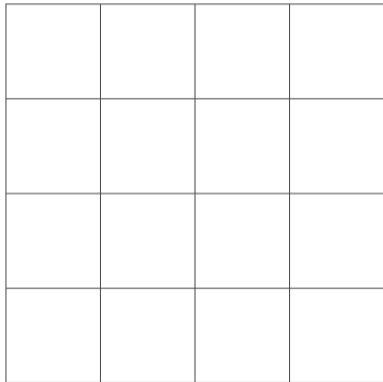
1. select the LR B-spline contributing more to the approximation error (in some sense),

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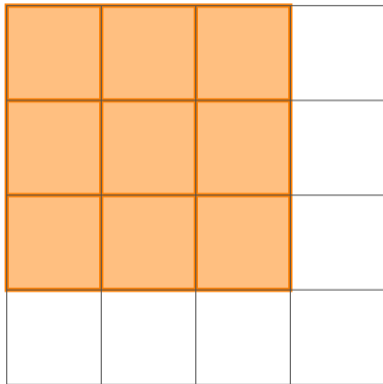
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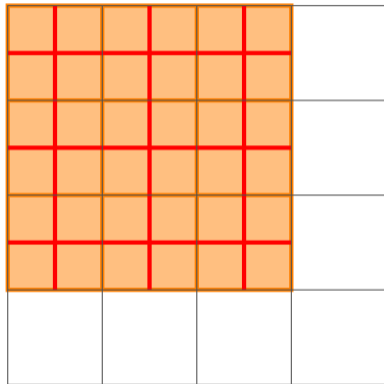
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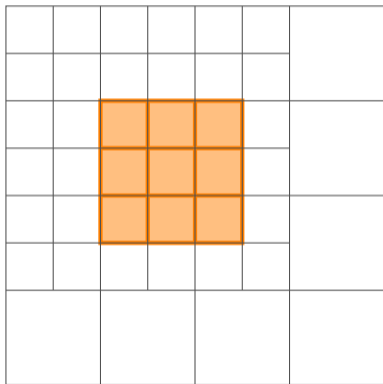
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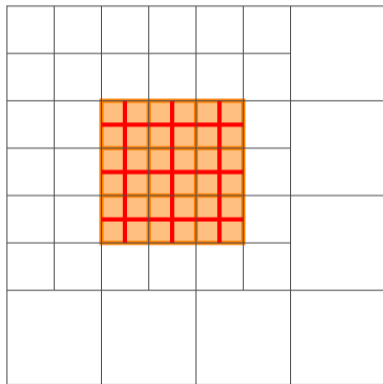
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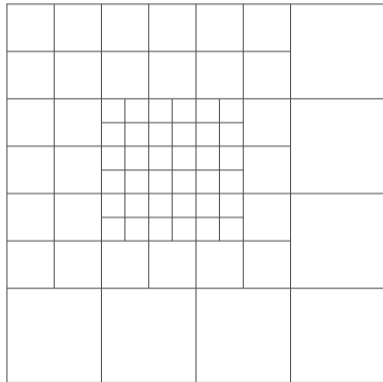
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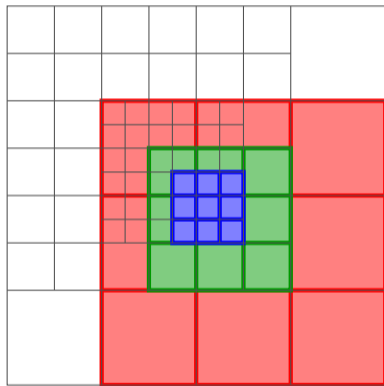
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final mesh

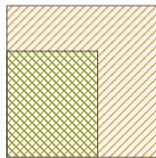
Non-Nested Support Structured (N_2S_2) Mesh Refinement

1. select the LR B-spline contributing more to the approximation error (in some sense),
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final mesh (no N_2S property)

recall, NOT OK:



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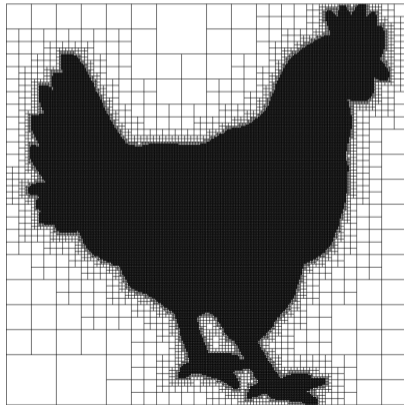
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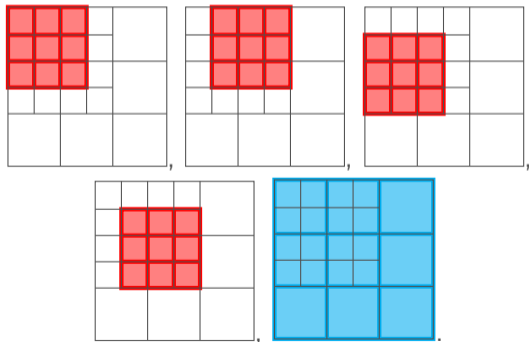
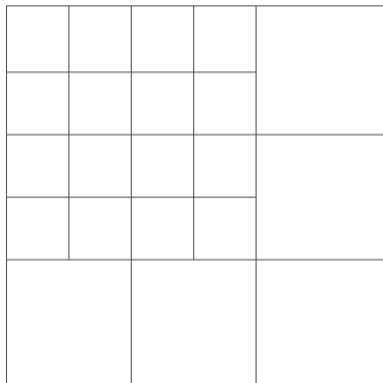


Non-Nested Support Structured (N_2S_2) Mesh Refinement

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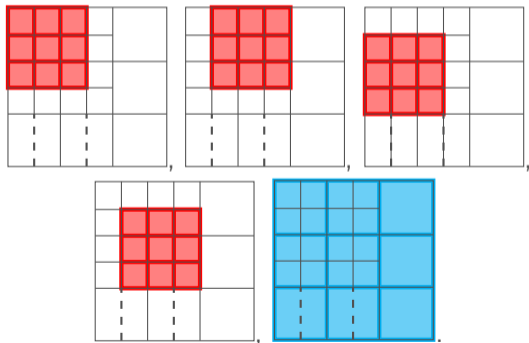
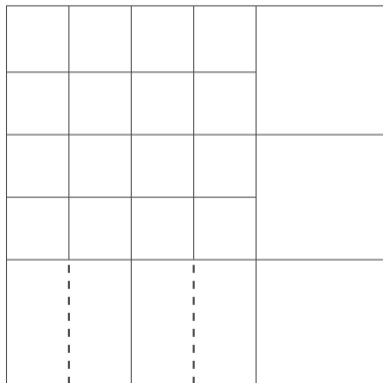
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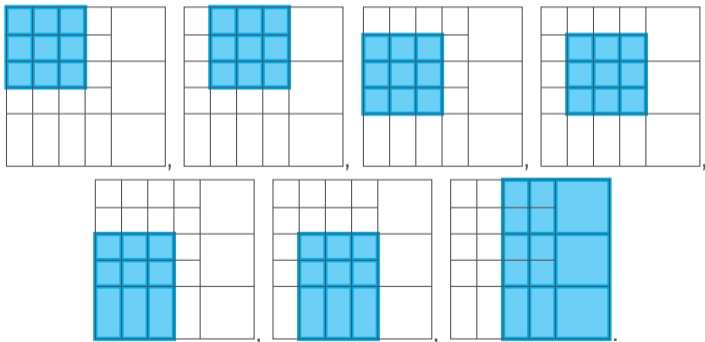
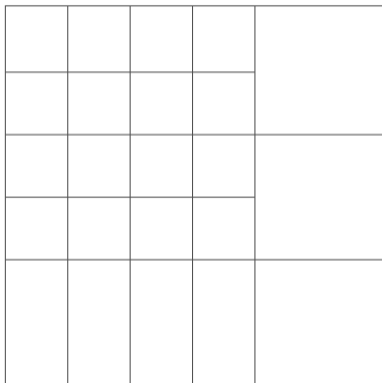
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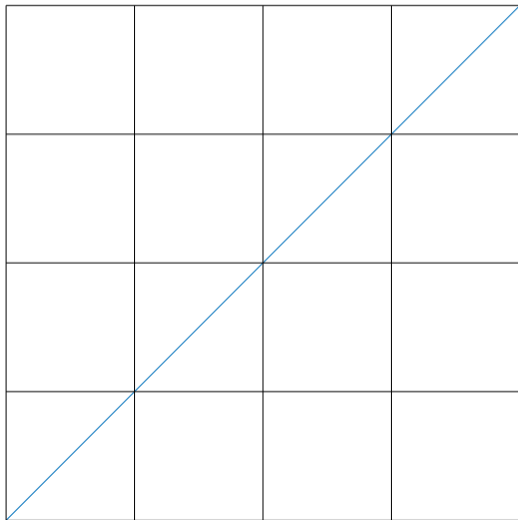
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3. modify the boundary of the region where the refinement is applied
4. apply the LR B-splines generation algorithm to refine the space.



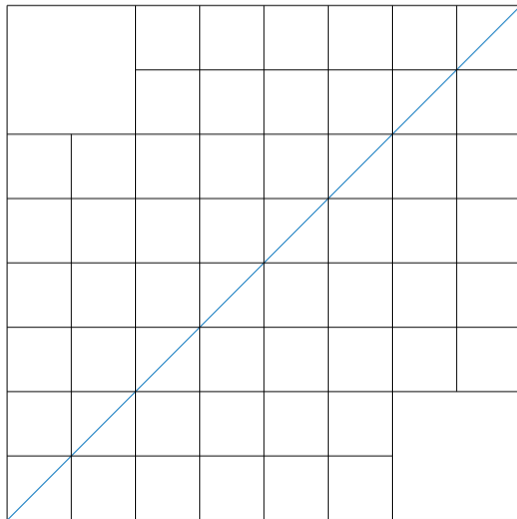
Non-Nested Support Structured (N_2S_2) Mesh Refinement

Example: degree $(2, 2)$.



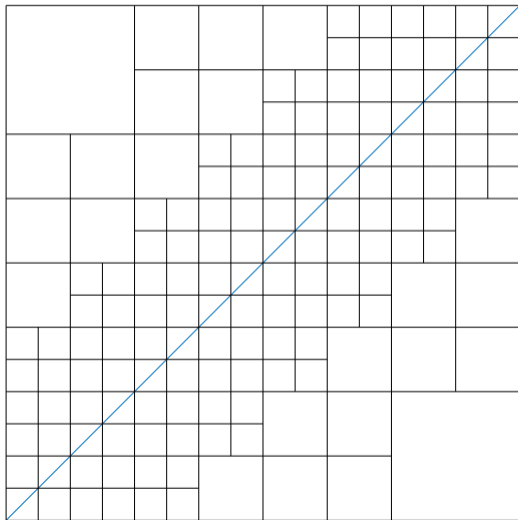
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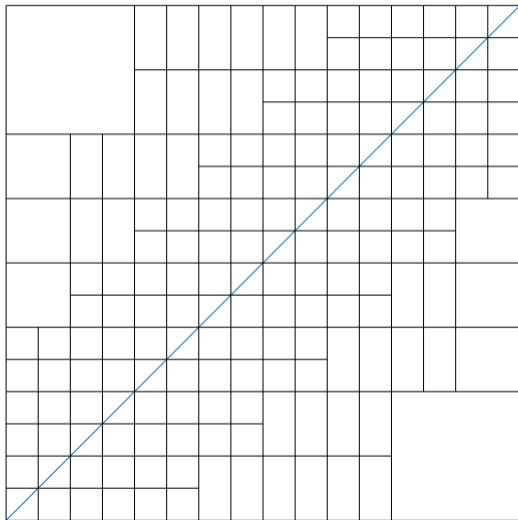
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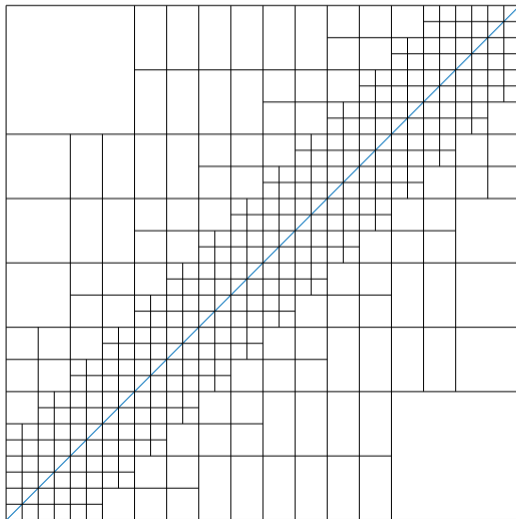
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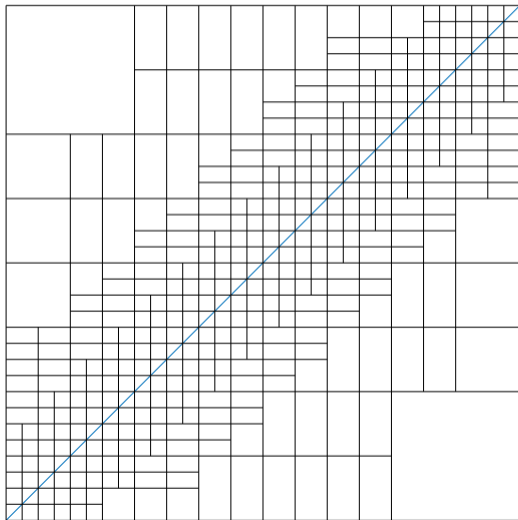
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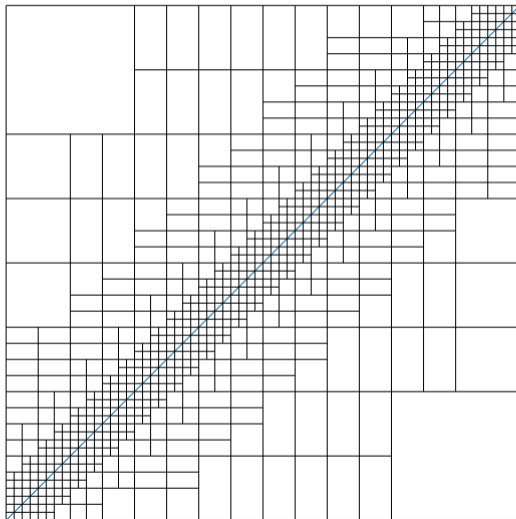
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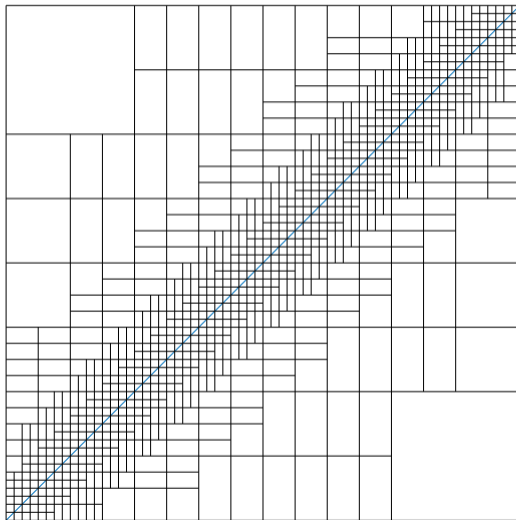
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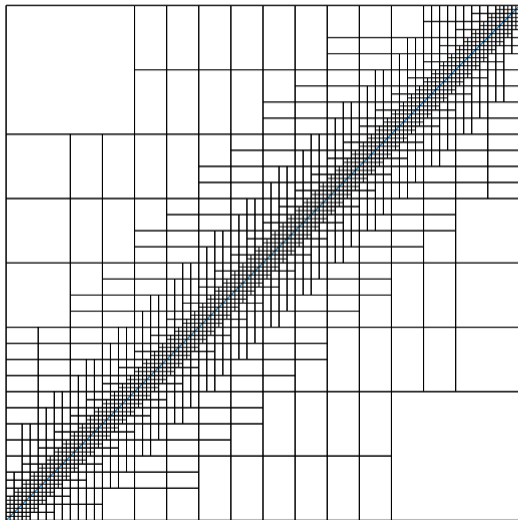
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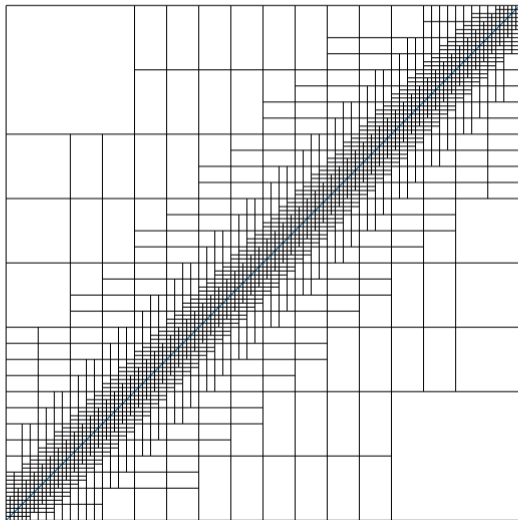
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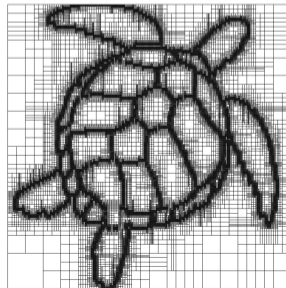
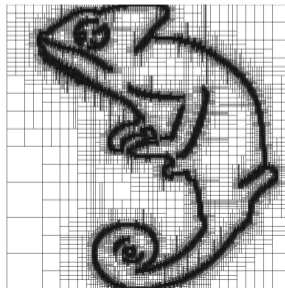
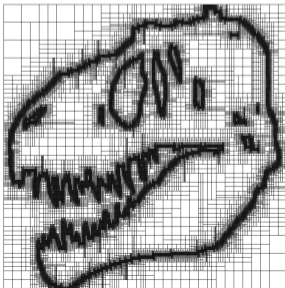
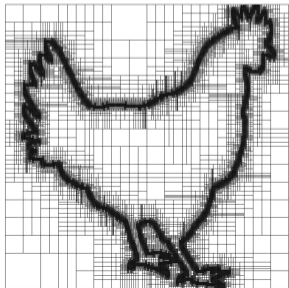
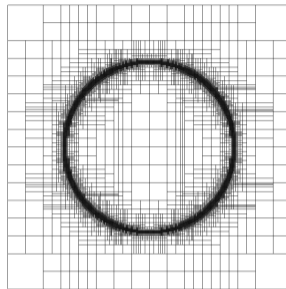
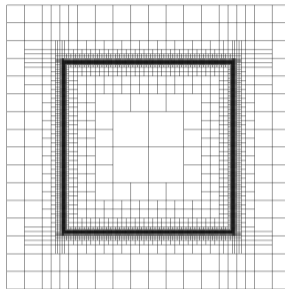
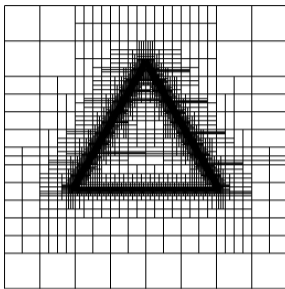
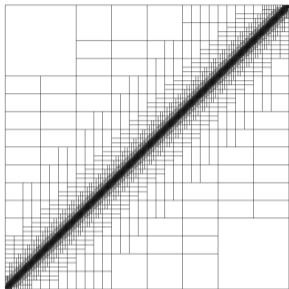


Non-Nested Support Structured (N_2S_2) Mesh Refinement

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Non-Nested Support Structured (N_2S_2) Mesh Refinement



Hierarchical LR-Mesh Costruction

Shadow Map:

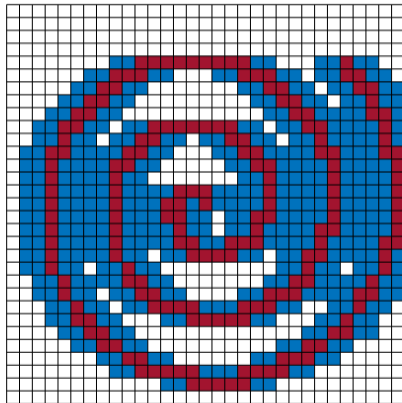
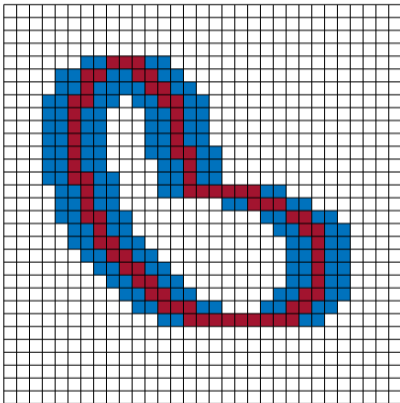


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Hierarchical LR-Mesh Costruction

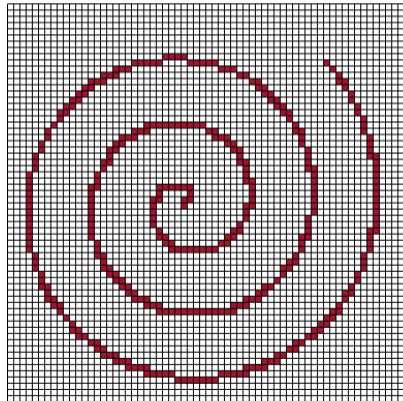
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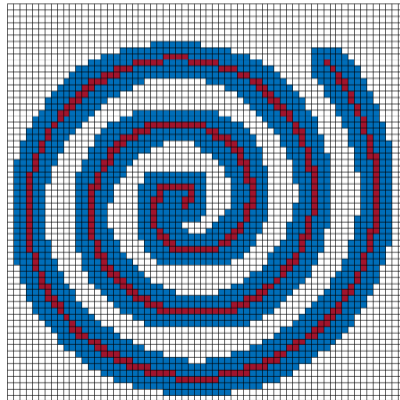
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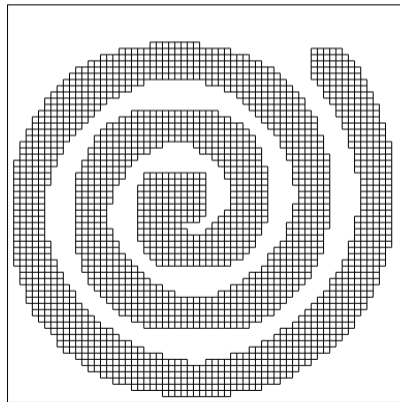
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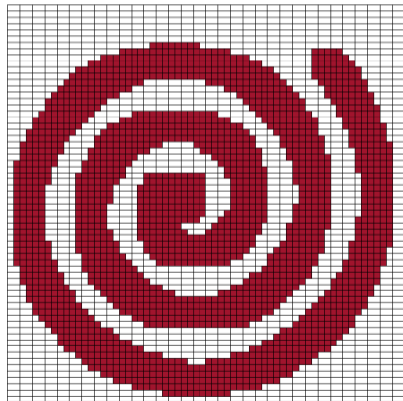
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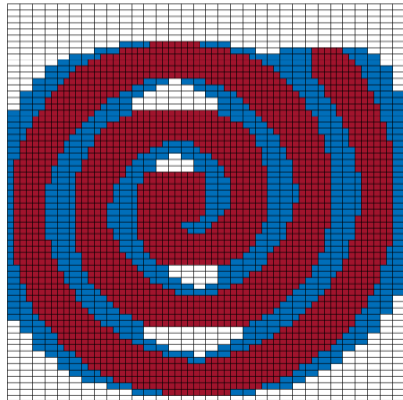
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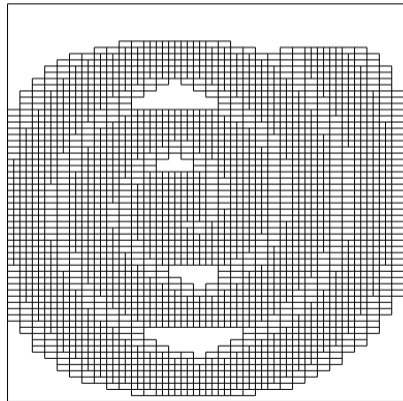
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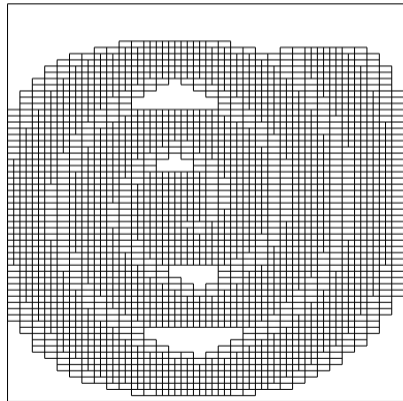
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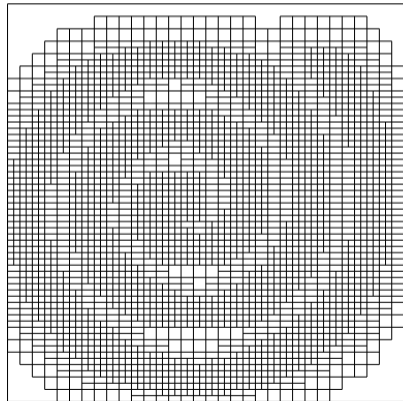
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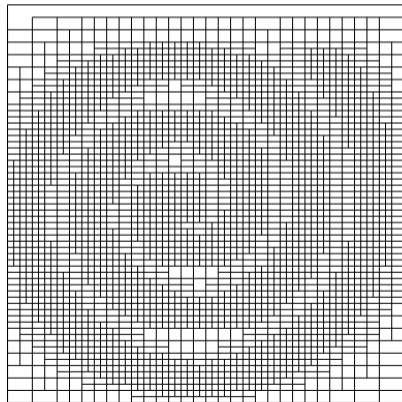
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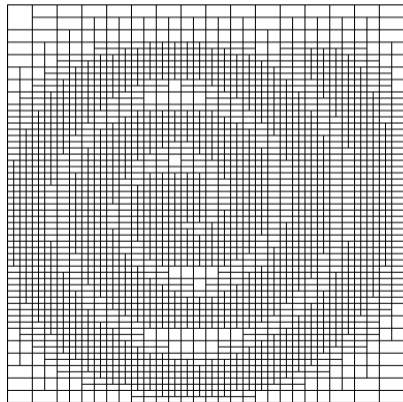
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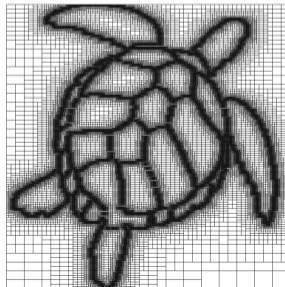
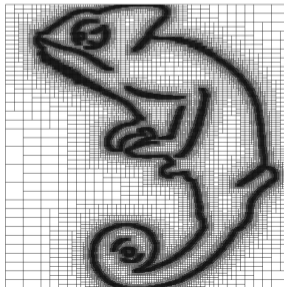
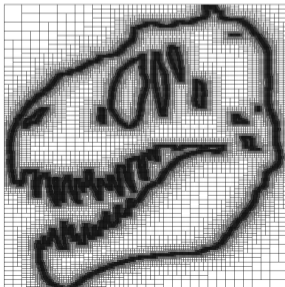
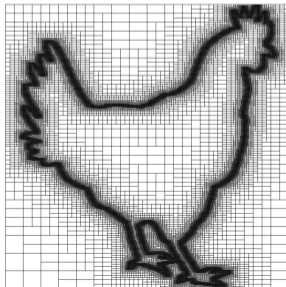
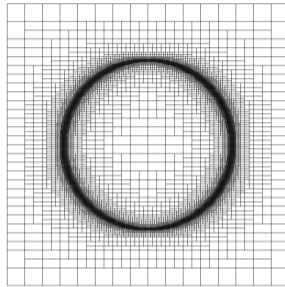
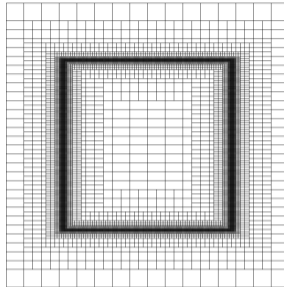
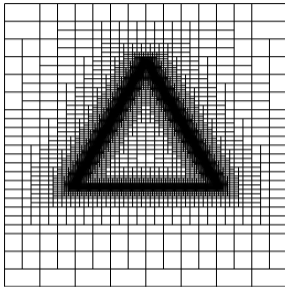
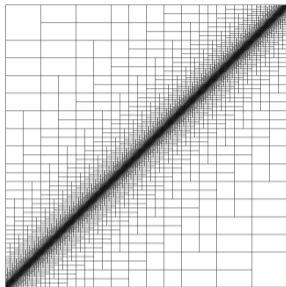
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Hierarchical LR-Mesh Costruction



Effective Grading Refinement Strategy



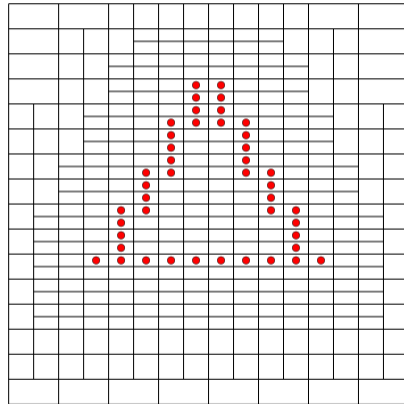
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Effective Grading Refinement Strategy



Refinement macro-step:

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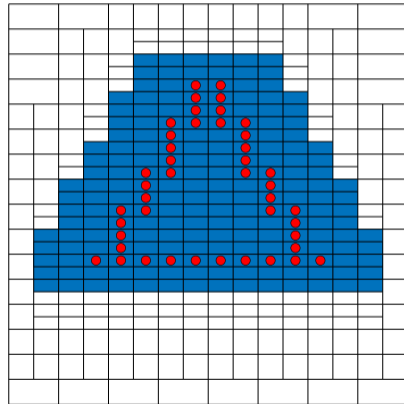


Effective Grading Refinement Strategy



Refinement macro-step:

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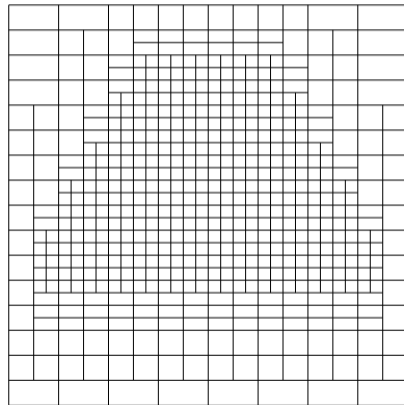


Effective Grading Refinement Strategy



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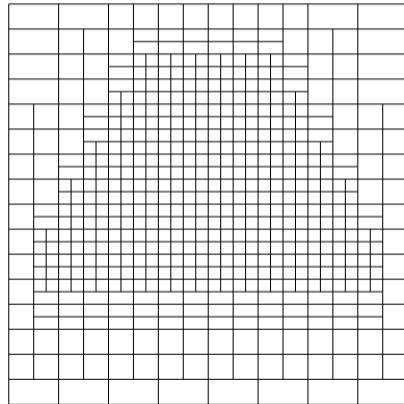
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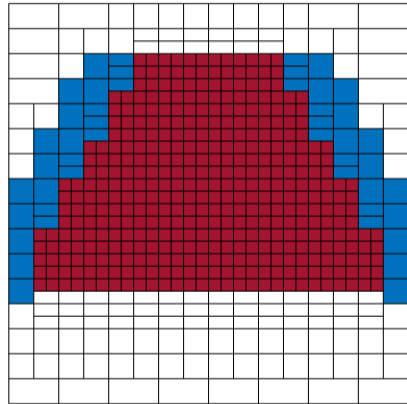
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N_2S reinstatement macro-step:

4. Consider the smallest boxes on the mesh and compute the shadow of such region (horizontal if square boxes, vertical if rectangular boxes),



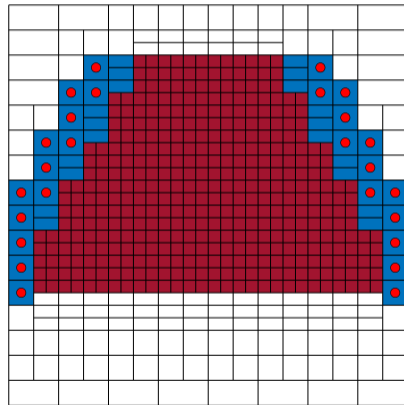
Effective Grading Refinement Strategy

Refinement macro-step:

1. Given a set of boxes marked for refinement,
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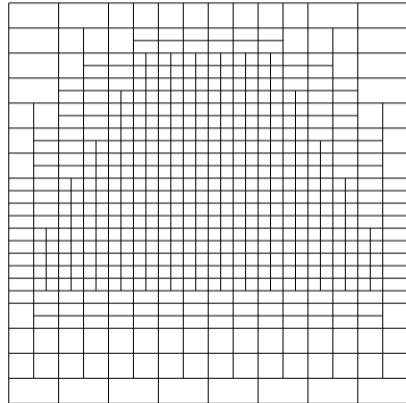
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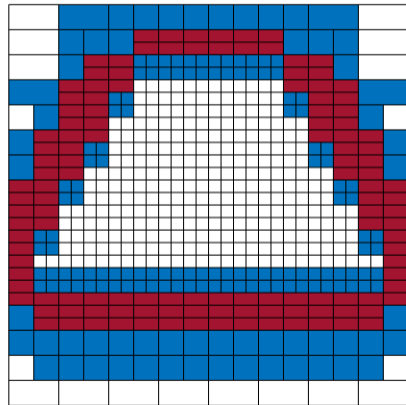
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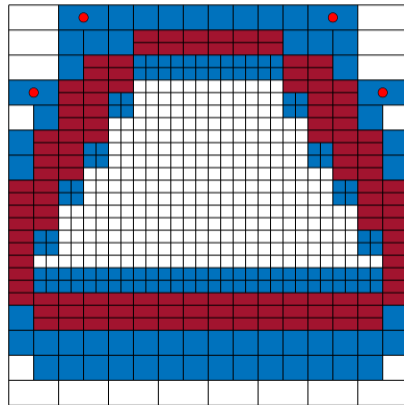
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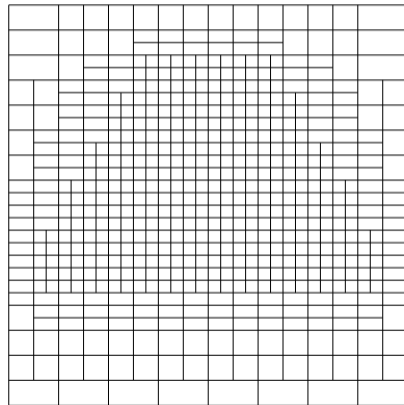
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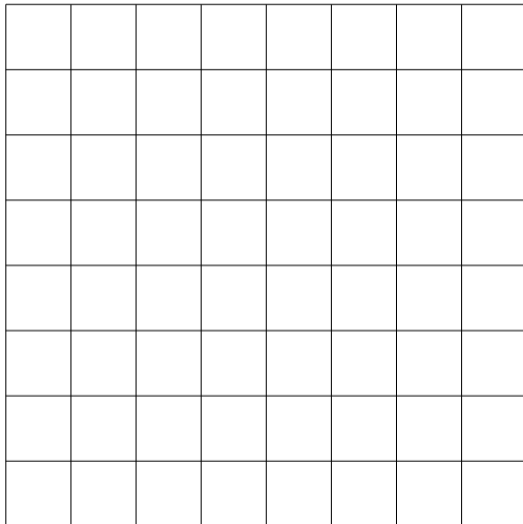
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Effective Grading Refinement Strategy

Example: degree (2,2), Triangle \rightarrow Circle \rightarrow Square

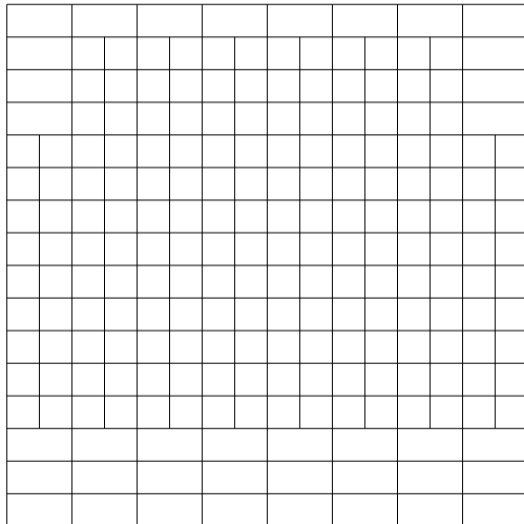


Effective Grading Refinement Strategy

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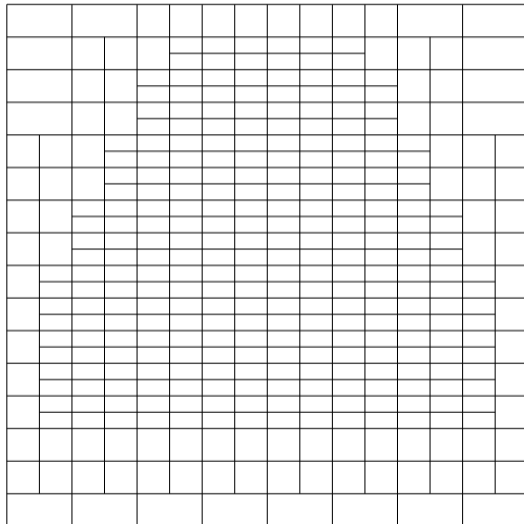
Effective Grading Refinement Strategy

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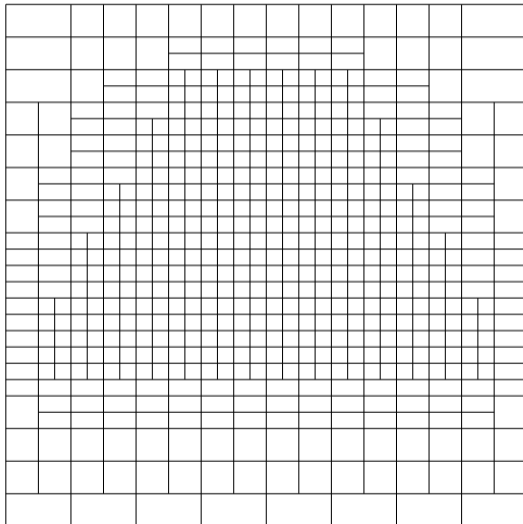
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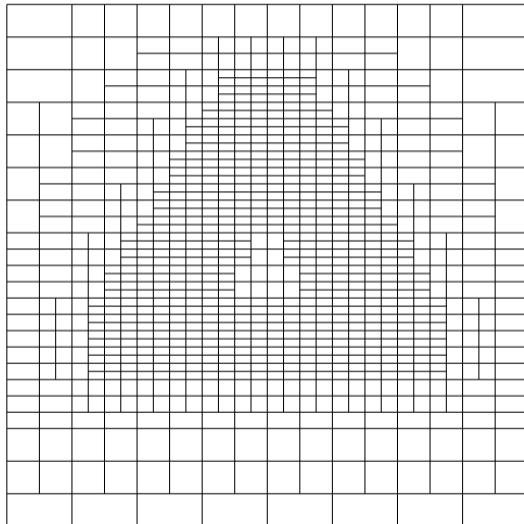
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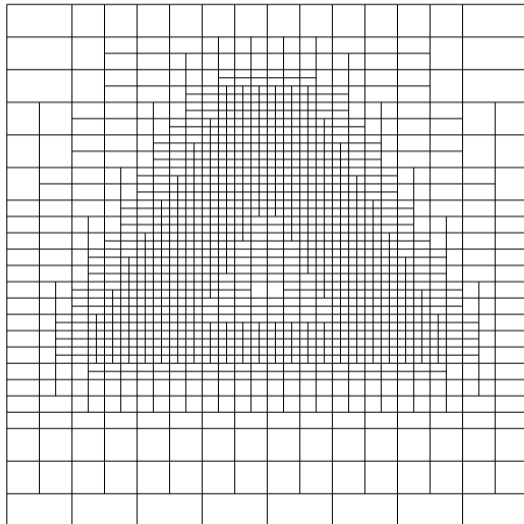
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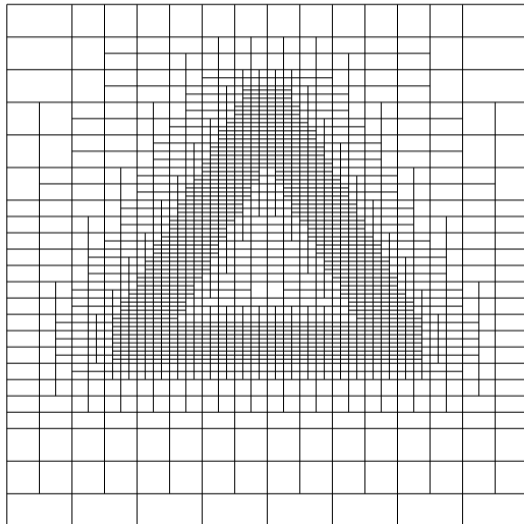
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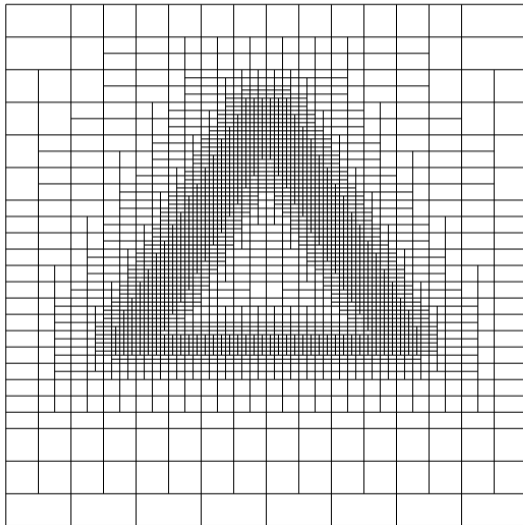
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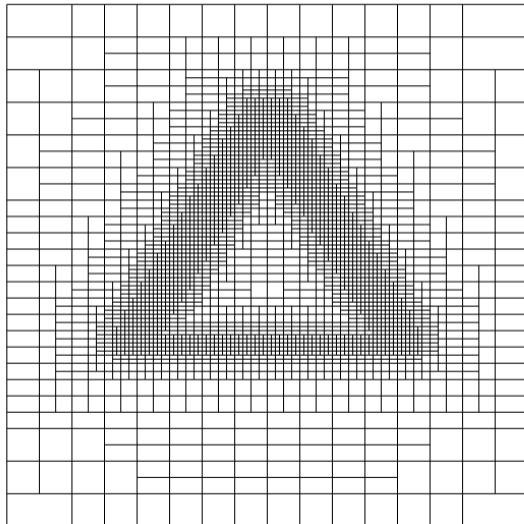
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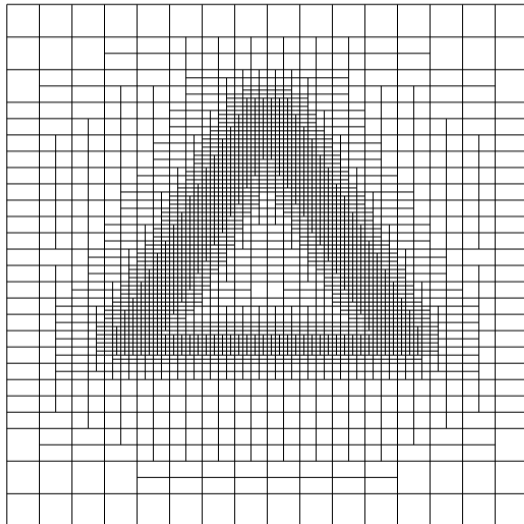
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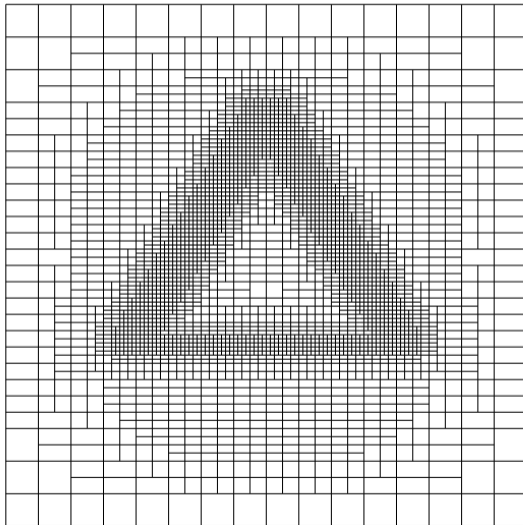
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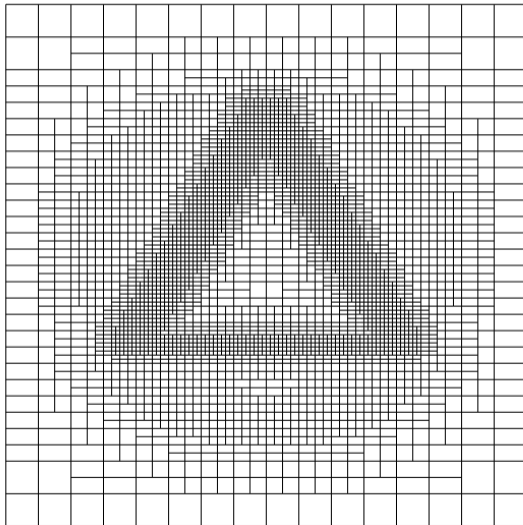
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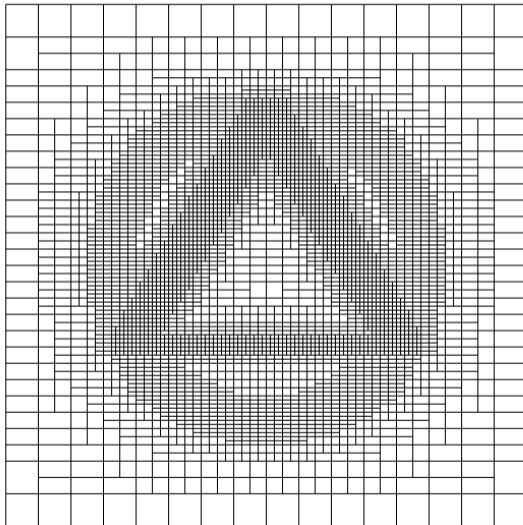
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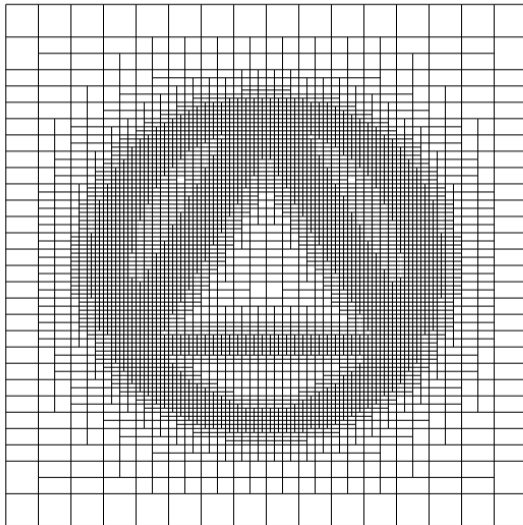
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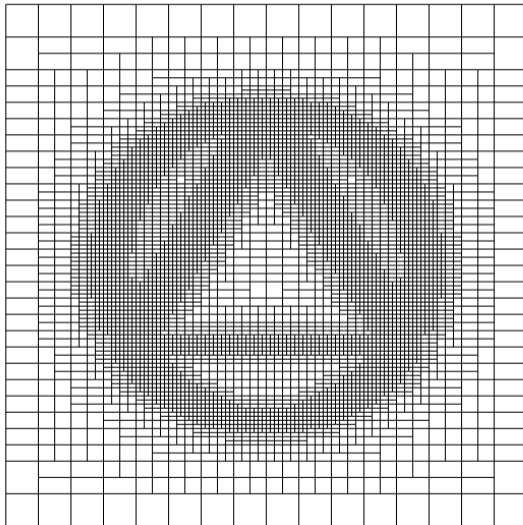
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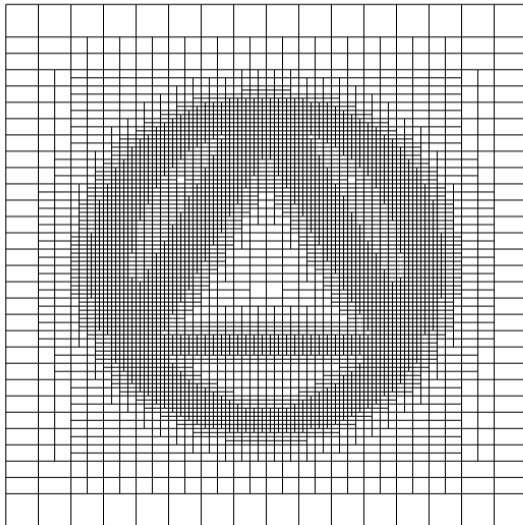
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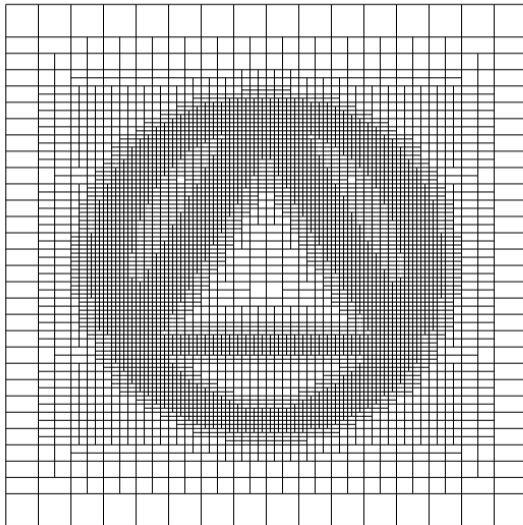
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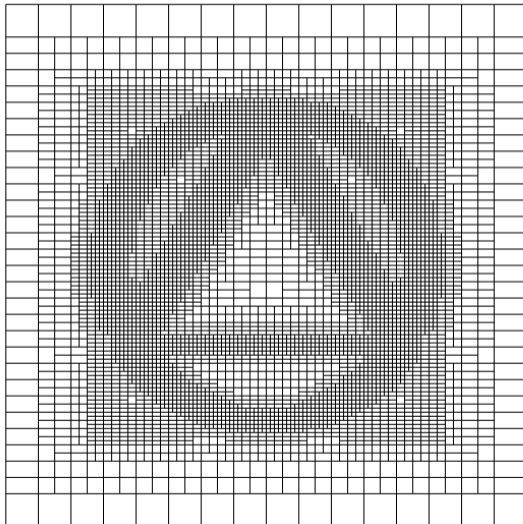
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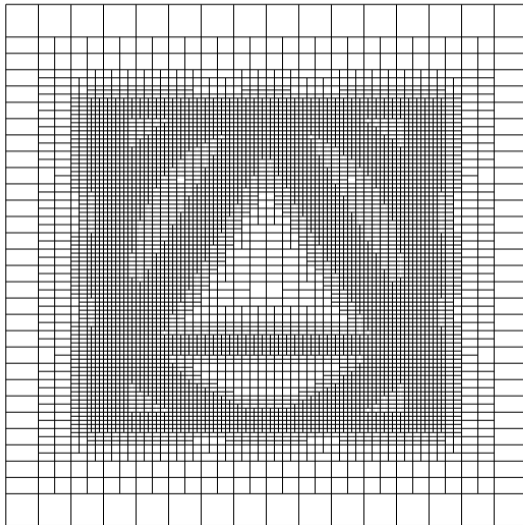
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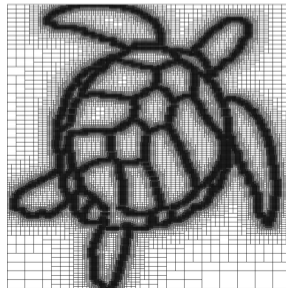
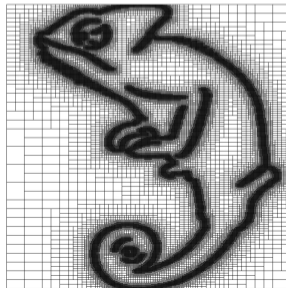
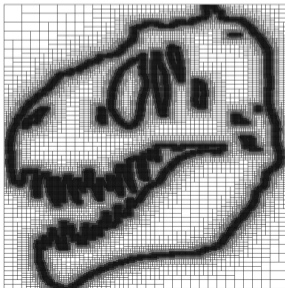
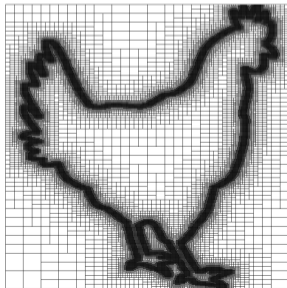
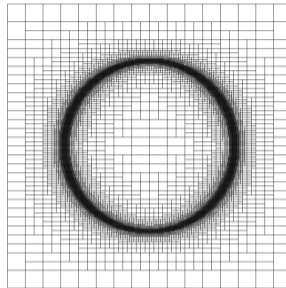
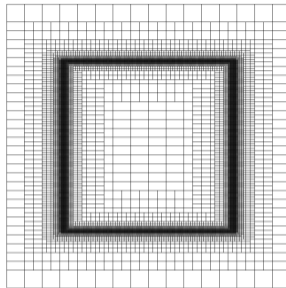
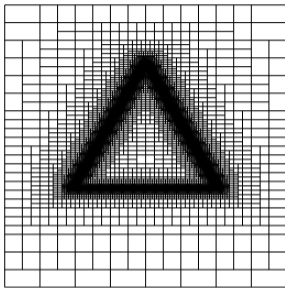
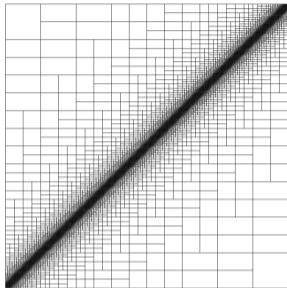


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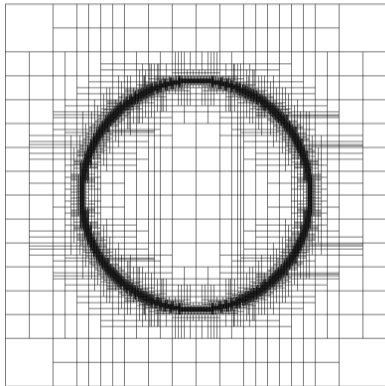
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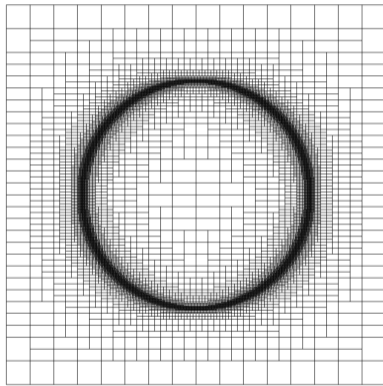
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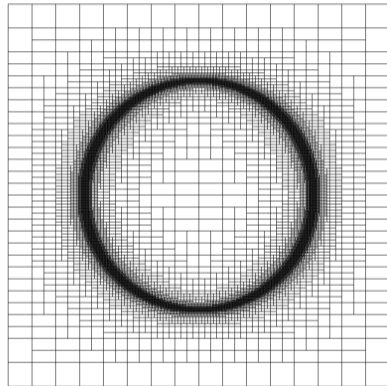
Comparison



N_2S_2 10818 LR B-splines



HLR 12374 LR B-splines



EG 15238 LR B-splines

One word on the spline space

\mathcal{N} mesh with local insertions (not necessarily LR mesh)

$$\mathbb{S}(\mathcal{N}) := \left\{ \begin{array}{l} f : \mathbb{R}^2 \rightarrow \mathbb{R} : \text{supp } f \subseteq \Omega, \\ f|_{\beta} \text{ is a polynomial of bidegree } \mathbf{p} \text{ in any } \beta \text{ box of } \mathcal{N}, \\ f \in C^{p_3-k-\mu(\gamma)}\text{-continuous across } \gamma \in \mathcal{N} \text{ in the } k\text{th direction.} \end{array} \right\}.$$

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Dimension formula: In general combinatorial part + homological part.

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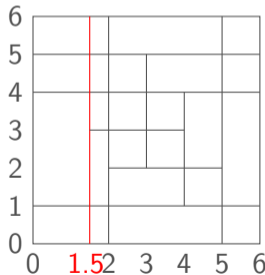
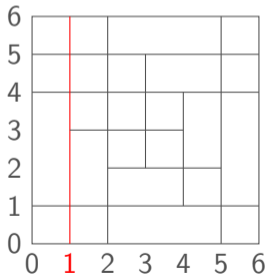
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$$\dim \mathbb{S}(\mathcal{N}) = 36$$



$$\dim \mathbb{S}(\mathcal{N}) = 36 + 1$$

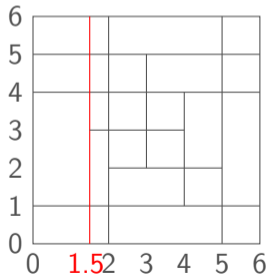
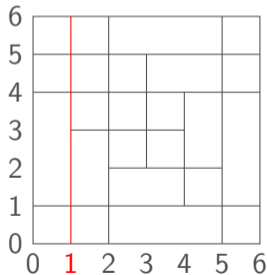
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On LR meshes only combinatorial ☺.

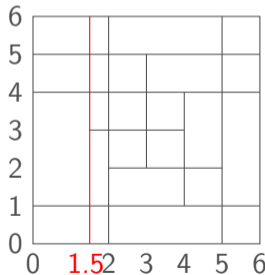
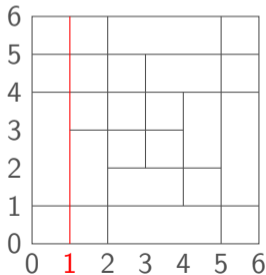
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$$\dim \mathbb{S}(\mathcal{N}) = 36 + 1$$

On LR meshes only combinatorial ☺. HB, THB, LR, ... $\subseteq \mathbb{S}(\mathcal{N})$.

Comparison

Adaptivity:



Comparison

Adaptivity: Local Refinement



Comparison

Adaptivity: Local Refinement + Change Region at any time

Comparison

Adaptivity: Local Refinement + Change Region at any time

Grading:

Comparison

Adaptivity: Local Refinement + Change Region at any time

Grading: Shape Regularity: No skinny elements



Comparison

Adaptivity: Local Refinement + Change Region at any time

Grading: Shape Regularity: No skinny elements + Local Quasi Uniformity:



no large boxes side by side small boxes

Comparison

Adaptivity: Local Refinement + Change Region at any time

Grading: Shape Regularity: No skinny elements + Local Quasi Uniformity:



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Completeness:

Comparison

Adaptivity: Local Refinement + Change Region at any time

Grading: Shape Regularity: No skinny elements + Local Quasi Uniformity:



no large boxes side by side small boxes

Completeness: The LR B-splines span the full ambient spline space $\mathcal{S}(\mathcal{N})$.

Comparison

Adaptivity: Local Refinement + Change Region at any time

Grading: Shape Regularity: No skinny elements + Local Quasi Uniformity:



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Marking:

Comparison

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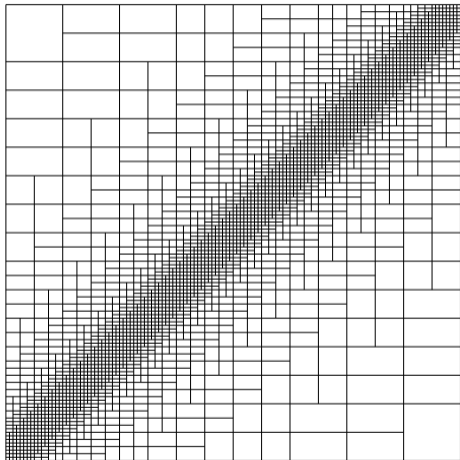
	(Loc.) Lin. Ind.	Adaptivity	Grading	Completeness	Marking
N_2S_2 strategy	✓	✓	✗	?	function-based
Hierarchical	Under Assumptions*	✗	✓	✓	box-based
Effective Grading	✓	✓	✓	✓	box-based

*fix maximal resolution and region of refinement *a priori*

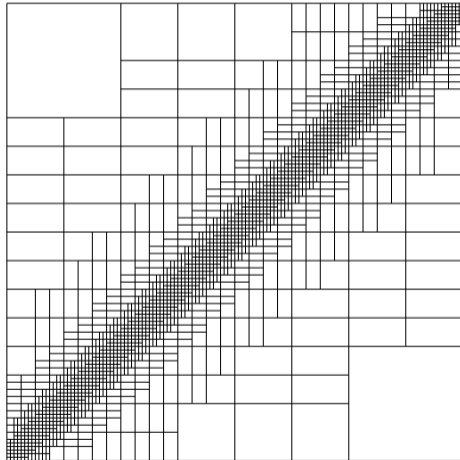
Conjecture: Local Linear Independence \Rightarrow Completeness.

Comparison of Adaptivity with and without Grading

From a refinement localized on a diagonal we switch to the other diagonal to form an “X”.



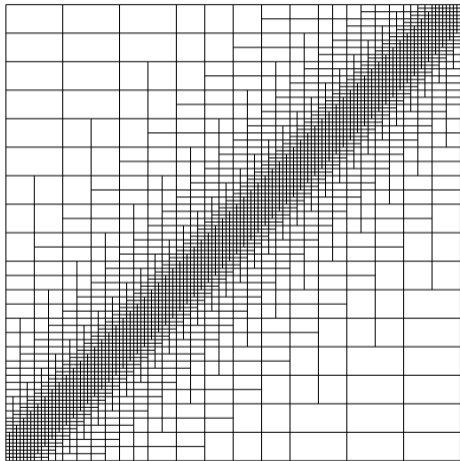
EG: Adaptivity & Grading



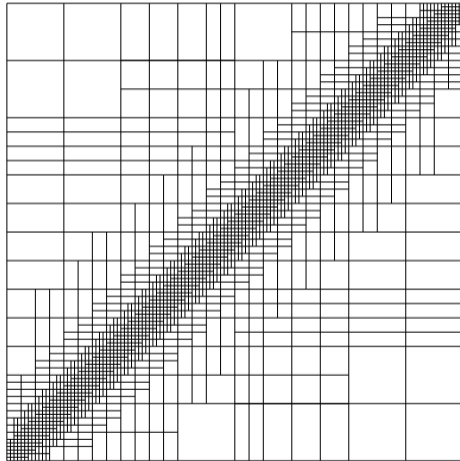
N_2S_2 : Adaptivity but no Grading

Comparison of Adaptivity with and without Grading

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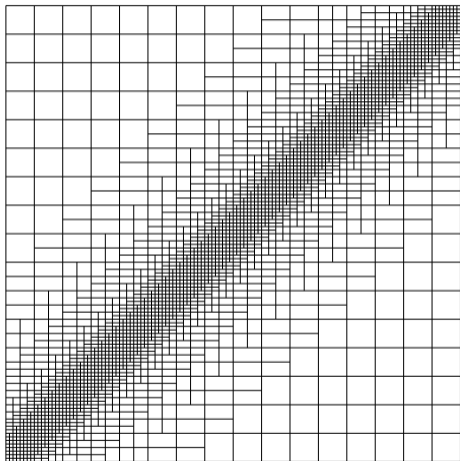
EG: Adaptivity & Grading



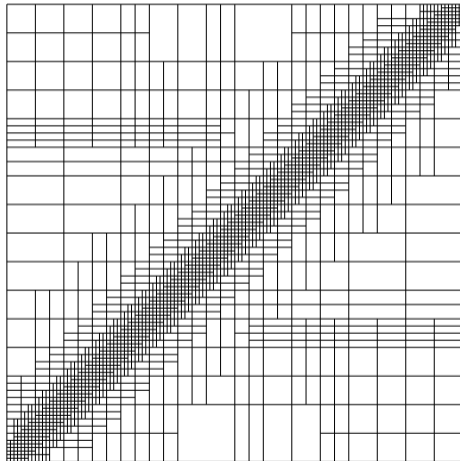
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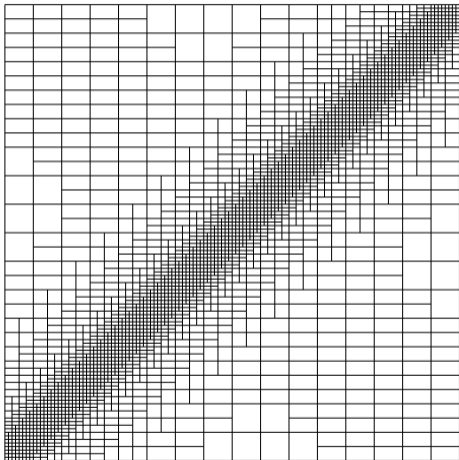
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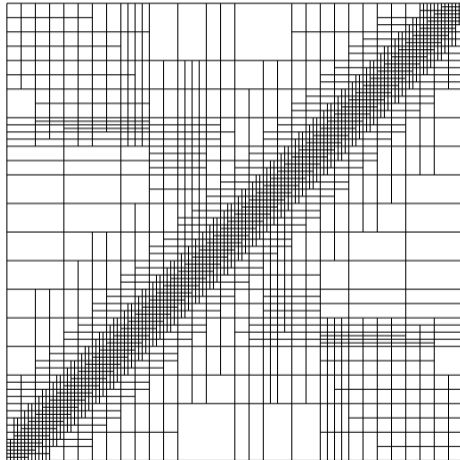
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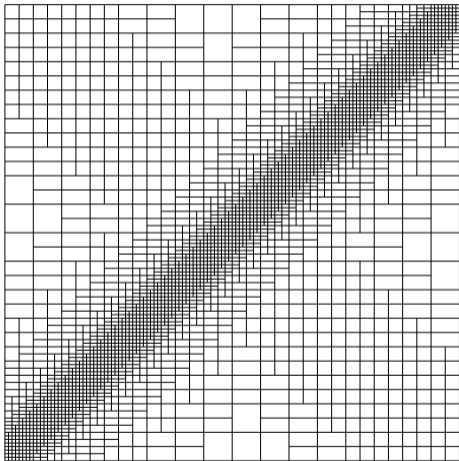
EG: Adaptivity & Grading



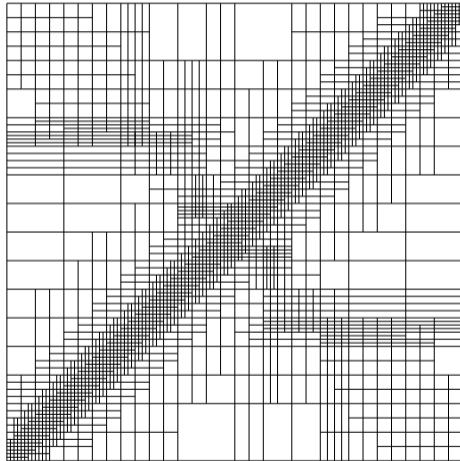
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Comparison of Adaptivity with and without Grading

From a refinement localized on a diagonal we switch to the other diagonal to form an “X”.



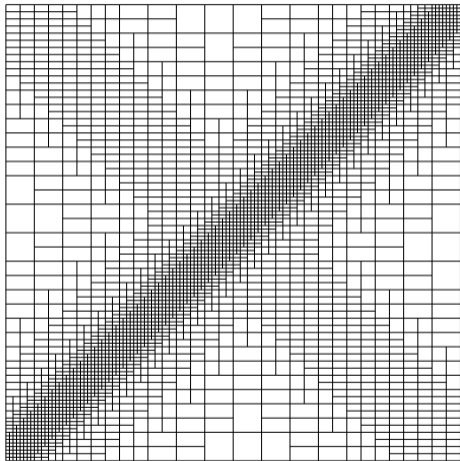
EG: Adaptivity & Grading



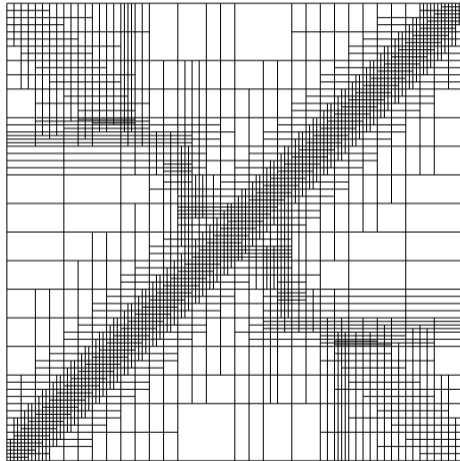
N_2S_2 : Adaptivity but no Grading

Comparison of Adaptivity with and without Grading

From a refinement localized on a diagonal we switch to the other diagonal to form an “X”.



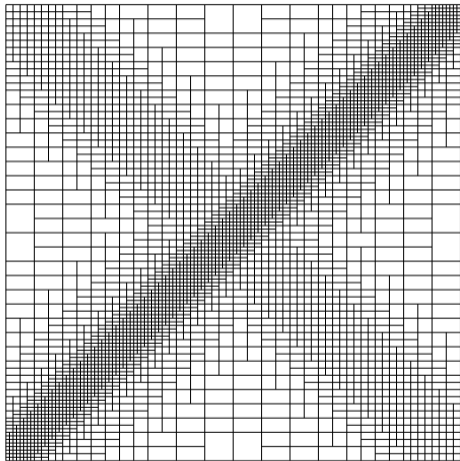
EG: Adaptivity & Grading



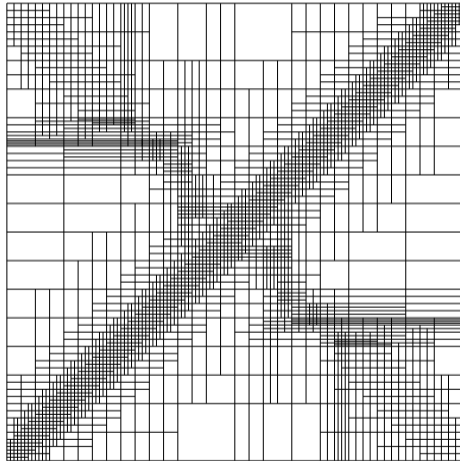
N_2S_2 : Adaptivity but no Grading

Comparison of Adaptivity with and without Grading

From a refinement localized on a diagonal we switch to the other diagonal to form an “X”.



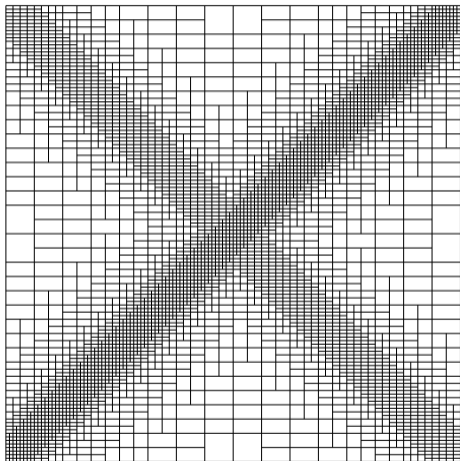
EG: Adaptivity & Grading



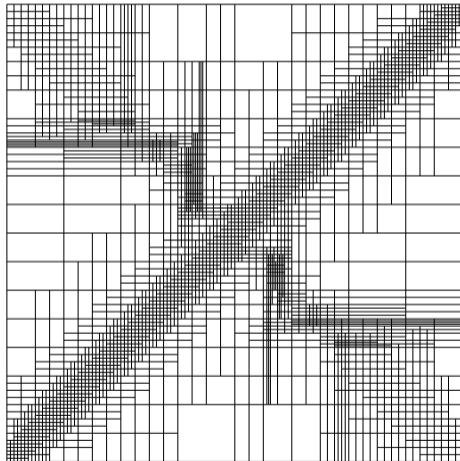
N_2S_2 : Adaptivity but no Grading

Comparison of Adaptivity with and without Grading

From a refinement localized on a diagonal we switch to the other diagonal to form an “X”.



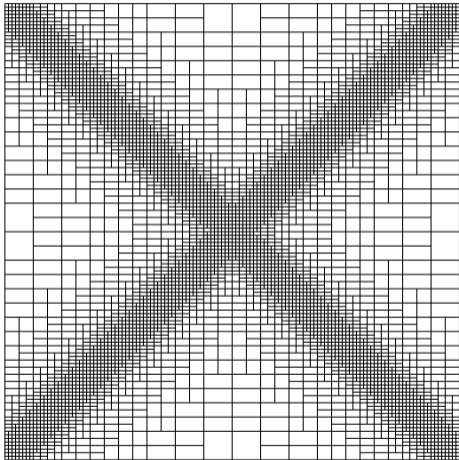
EG: Adaptivity & Grading



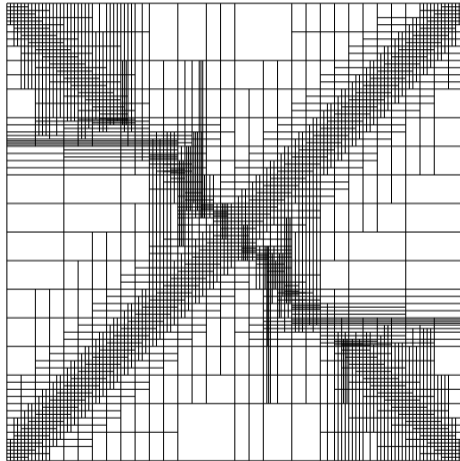
N_2S_2 : Adaptivity but no Grading

Comparison of Adaptivity with and without Grading

From a refinement localized on a diagonal we switch to the other diagonal to form an “X”.










EG: Adaptivity & Grading



N_2S_2 : Adaptivity but no Grading

References

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