

# Artificial Intelligence

AI2002

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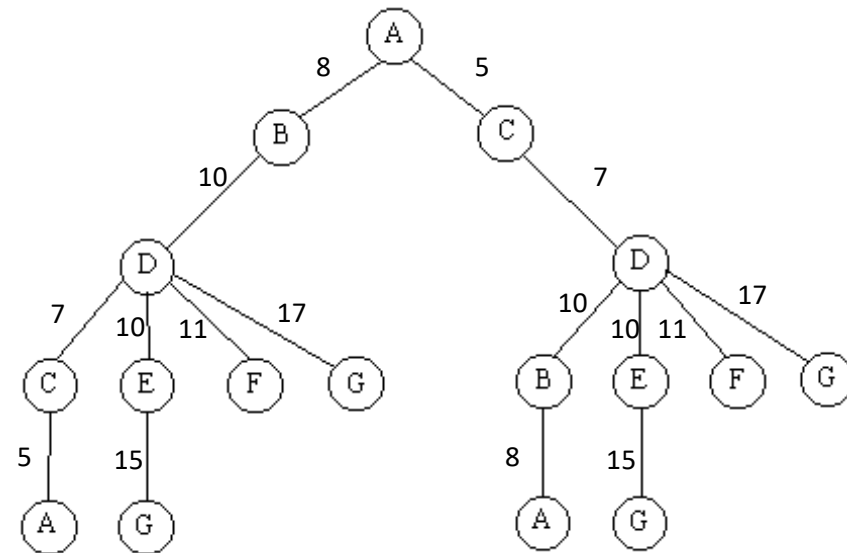
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# Uniform Cost Search

- Expands the node  $n$  with the lowest path cost  $g(n)$ .
- This is done by storing the frontier as a priority queue ordered by  $g$ .
- It explores cumulative path cost instead of heuristics that direct towards a goal making it a weighted version of BFS –an uninformed search.

# UCS (Priority Queue)

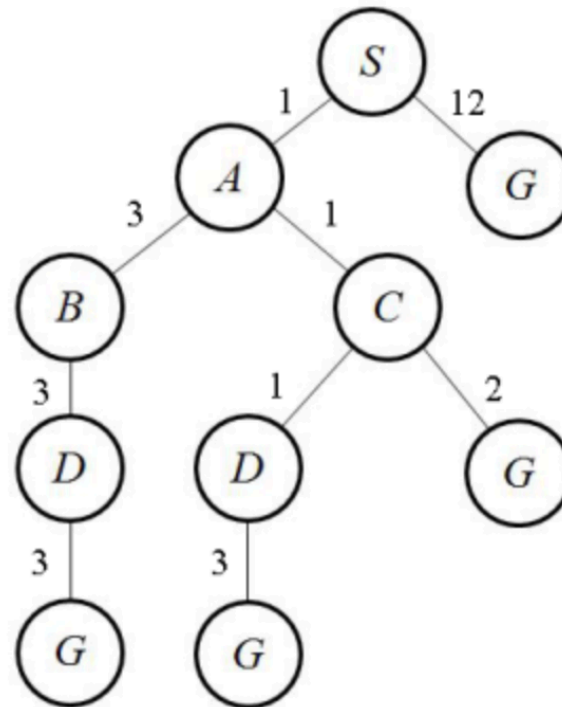
- begin
  - Open = [Start (path cost=0)];
  - Closed = [];
  - While open <> [] do
  - begin
    - remove highest priority (lowest path cost) state from open, call it X;
    - if X is a goal then return(success)
    - else begin
      - generate children of X;
      - put X on closed;
      - add the children of X if not in open or closed
      - replace the children of X if they already exist in open with higher path cost
  - end
  - return(failure)
- end.



1. Open = [A(0)]; closed = [ ];
2. Open = [B(8),C(5)]; closed = [A(0)];
3. Open = [B(8),D(12)]; closed = [C(5),A(0)];
4. Open = [D(12)]; closed = [B(8),C(5),A(0)];
5. Open = [E(22),F(23),G(29)]; closed = [D(12),B(8),C(5),A(0)];
6. Open = [F(23),G(29)]; closed = [E(22),D(12),B(8),C(5),A(0)];
7. Open = [G(29)]; closed = [F(23), E(22),D(12),B(8),C(5),A(0)];

# Activity

- Perform Uniform cost search on the given tree
- Show how open and closed queues change in each iteration



# Depth Limited Search

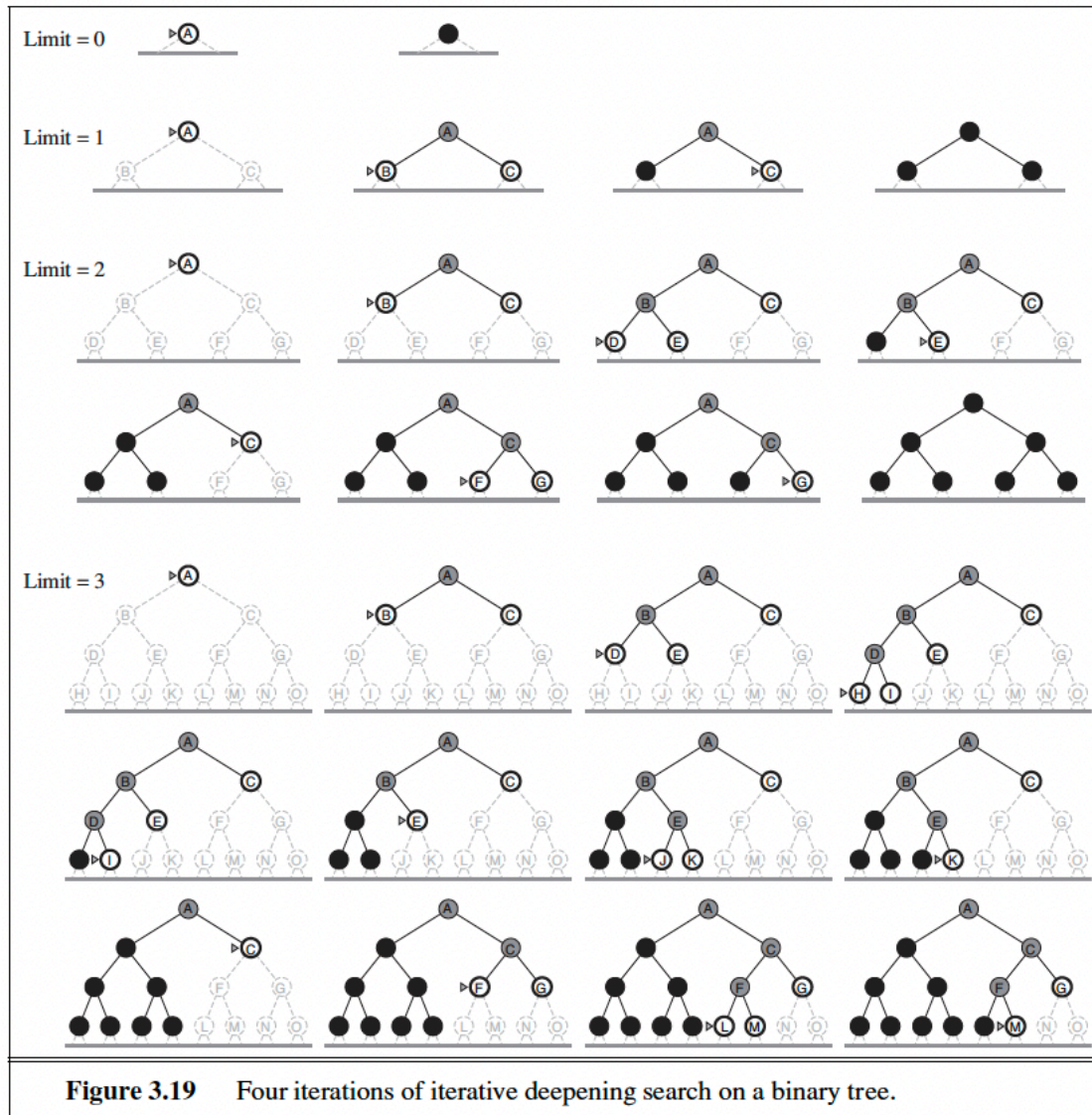
- The failure of depth-first search in infinite state spaces can be alleviated by supplying depth-first search with a predetermined depth limit.
- The nodes at depth are treated as if they have no successors.
- The depth limit solves the infinite-path problem.

# Iterative Deepening Search

- This form of search is an **excellent compromise** between depth-first and breadth-first searches.
- The search **sets a depth limit** and then does a **depth-first search down** to this **limit** - if a solution is **found** it exits with **success**.
- If a solution **is not found** then the **limit is increased** and the search run again but down to this new depth.
- Obviously, a large number of nodes are **revisited a number of times**, but the **overall loss due to this becomes insignificant** for a search through a **large** tree.
- The **number of visited nodes grows** exponentially as you go down the tree so almost all the time is spent at **the deepest levels** which are **only visited once** unless the **search repeats to a deeper level**, at which the time spent in these levels will become insignificant.
- The memory consumption is the **same** as that for a **depth-first search**, but the search is always guaranteed to find a solution, as in **breadth-first search**.

# IDS

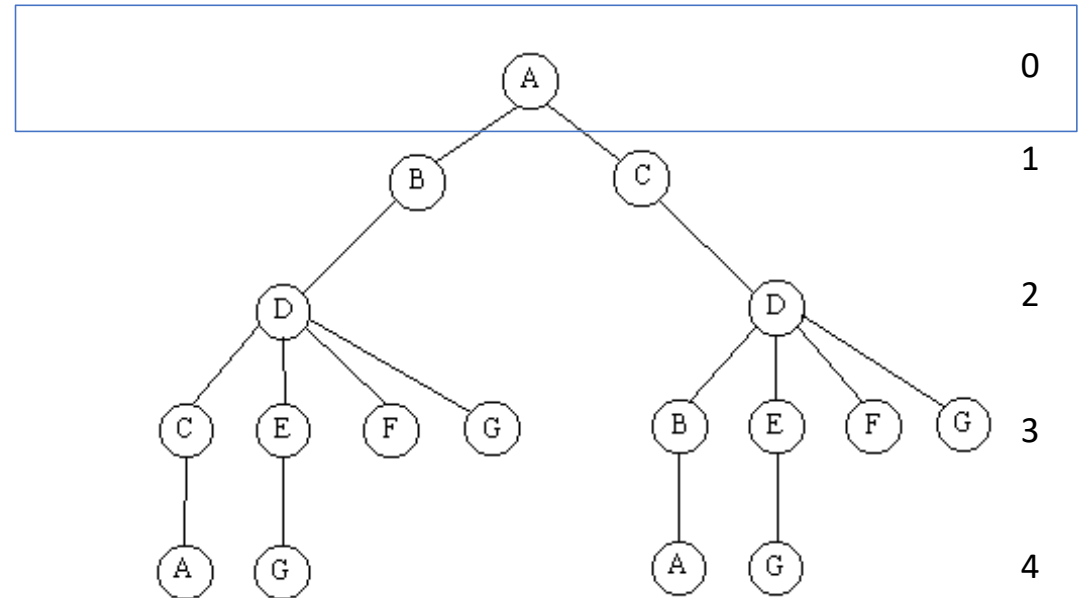
- Iterative deepening search (or iterative deepening depth-first search) is a general strategy, often used in combination with depth-first tree search, that finds the best depth limit.
- It does this by gradually increasing the limit—first 0, then 1, then 2, and so on—until a goal is found.
- This will occur when the depth limit reaches  $d$ , the depth of the shallowest goal node.
- Iterative deepening combines the benefits of depth-first and breadth-first search.





# IDS

- begin
- for limit = 0 to  $\infty$  do
- begin
- Open = [(Start, 0)];
- Closed = [];
- while Open  $\neq$  [] do
- begin
- remove leftmost state (X, d) from Open;
- if X is a goal then
- return(success);
- else if d < limit then
- begin
- generate children of X;
- put X on Closed;
- discard children of X if already on Open or Closed;
- put remaining children on left end of Open with depth d+1;
- end
- end
- end
- return(failure)
- end

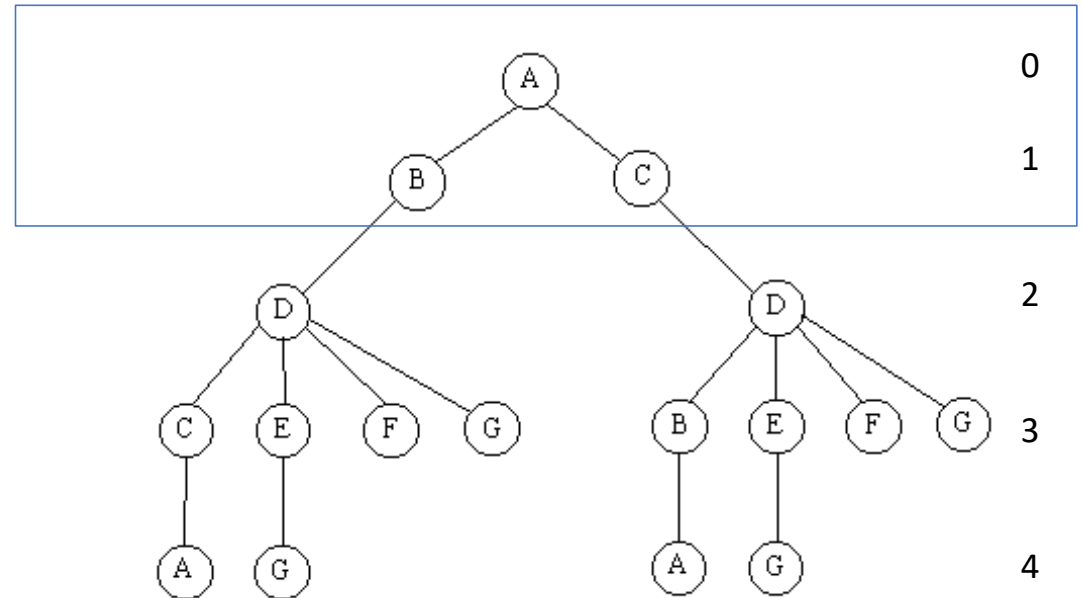


Iteration 1:

1. Open = [A]; closed = [ ];
2. Open = [ ] ; closed = [A];

# IDS

- begin
- for limit = 0 to  $\infty$  do
- begin
- Open = [(Start, 0)];
- Closed = [];
- while Open  $\neq$  [] do
- begin
- remove leftmost state (X, d) from Open;
- if X is a goal then
- return(success);
- else if d < limit then
- begin
- generate children of X;
- put X on Closed;
- discard children of X if already on Open or Closed;
- put remaining children on left end of Open with depth d+1;
- end
- end
- end
- return(failure)
- end

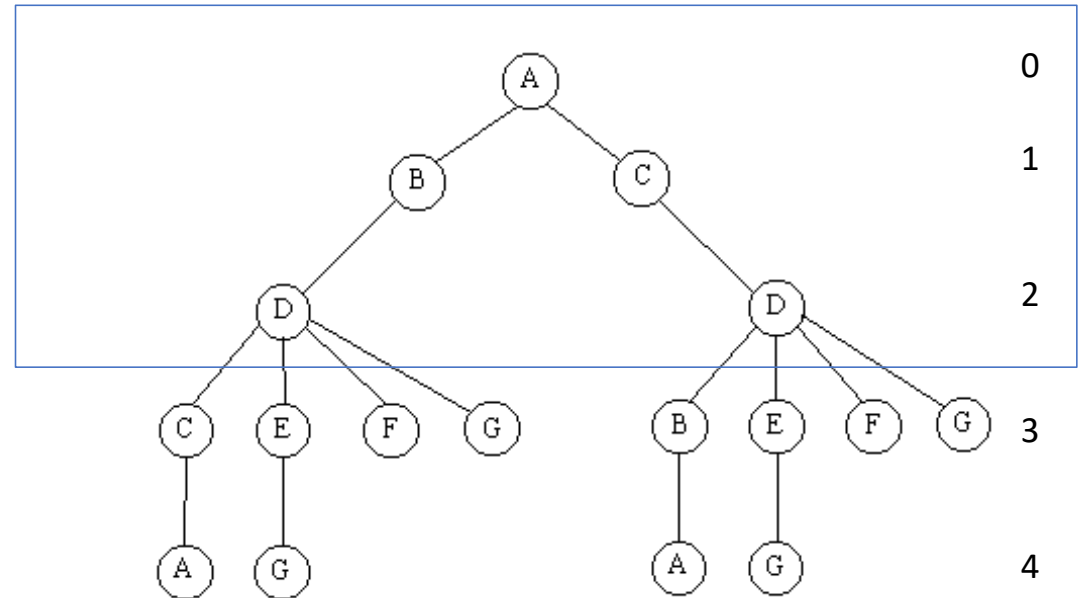


Iteration 2:

1. Open = [A]; closed = [ ];
2. Open = [B,C]; closed = [A];
3. Open = [C]; closed = [B,A];
4. Open = [ ]; closed = [C,B,A];

# IDS

- begin
- for limit = 0 to  $\infty$  do
- begin
- Open = [(Start, 0)];
- Closed = [];
- while Open  $\neq$  [] do
- begin
- remove leftmost state (X, d) from Open;
- if X is a goal then
- return(success);
- else if d < limit then
- begin
- generate children of X;
- put X on Closed;
- discard children of X if already on Open or Closed;
- put remaining children on left end of Open with depth d+1;
- end
- end
- end
- return(failure)
- end

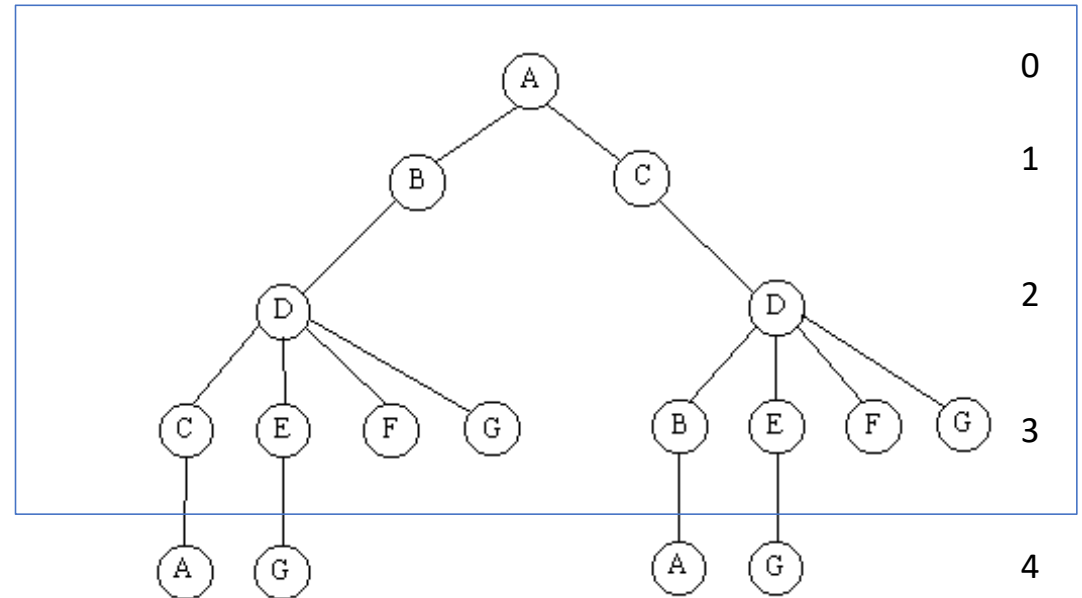


Iteration 3:

1. Open = [A]; closed = [ ];
2. Open = [B,C]; closed = [A];
3. Open = [D,C]; closed = [B,A];
4. Open = [C]; closed = [D,B,A];
5. Open = []; closed = [C,D,B,A];

# IDS

- begin
- for limit = 0 to  $\infty$  do
- begin
- Open = [(Start, 0)];
- Closed = [];
- while Open  $\neq$  [] do
- begin
- remove leftmost state (X, d) from Open;
- if X is a goal then
- return(success);
- else if d < limit then
- begin
- generate children of X;
- put X on Closed;
- discard children of X if already on Open or Closed;
- put remaining children on left end of Open with depth d+1;
- end
- end
- end
- return(failure)
- end

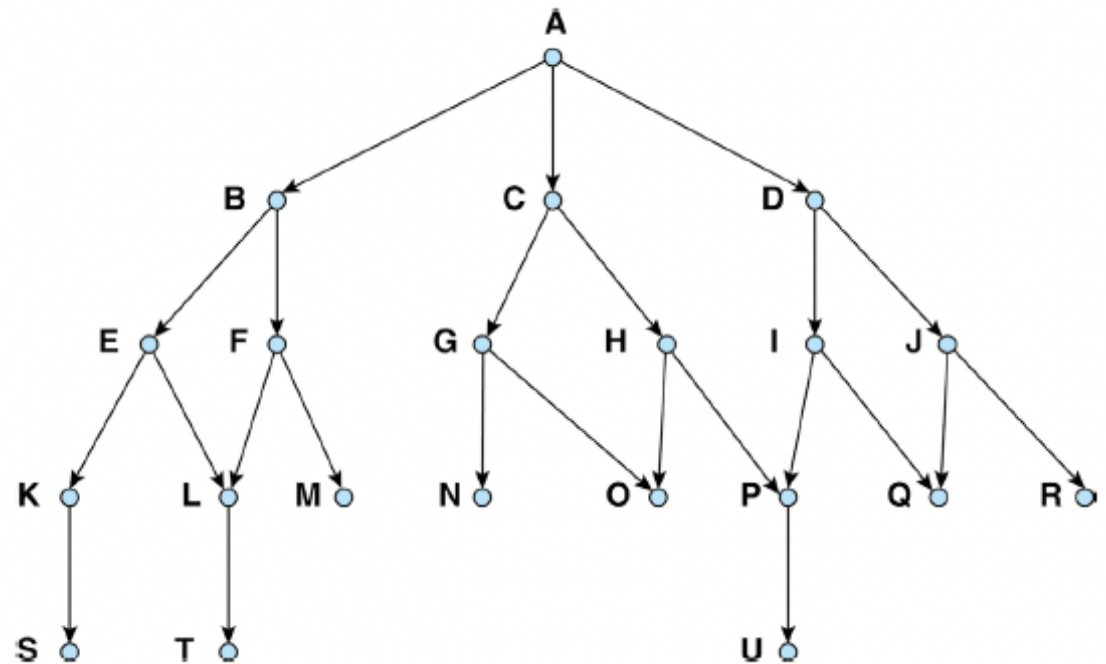


Iteration 4:

1. Open = [A]; closed = [ ];
2. Open = [B,C]; closed = [A];
3. Open = [D,C]; closed = [B,A];
4. Open = [E,F,G,C]; closed = [D,B,A];
5. Open = [F,G,C]; closed = [E,D,B,A];
6. Open = [G,C]; closed = [F,E,D,B,A];

# Activity

- Perform IDS on the given tree from A to G
- Write down how each iteration is reflected in open and closed queues



# Comparing uninformed search strategies

- For graph searches, the main differences are that depth-first search is complete for finite state spaces and that the space and time complexities are bounded by the size of the state space.

Criterion	Breadth-First	Uniform-Cost	Depth-First	Depth-Limited	Iterative Deepening	Bidirectional (if applicable)
Complete?	Yes <sup>a</sup>	Yes <sup>a,b</sup>	No	No	Yes <sup>a</sup>	Yes <sup>a,d</sup>
Time	$O(b^d)$	$O(b^{1+\lceil C^*/\epsilon \rceil})$	$O(b^m)$	$O(b^l)$	$O(b^d)$	$O(b^{d/2})$
Space	$O(b^d)$	$O(b^{1+\lceil C^*/\epsilon \rceil})$	$O(bm)$	$O(bl)$	$O(bd)$	$O(b^{d/2})$
Optimal?	Yes <sup>c</sup>	Yes	No	No	Yes <sup>c</sup>	Yes <sup>c,d</sup>

**Figure 3.21** Evaluation of tree-search strategies.  $b$  is the branching factor;  $d$  is the depth of the shallowest solution;  $m$  is the maximum depth of the search tree;  $l$  is the depth limit. Superscript caveats are as follows: <sup>a</sup> complete if  $b$  is finite; <sup>b</sup> complete if step costs  $\geq \epsilon$  for positive  $\epsilon$ ; <sup>c</sup> optimal if step costs are all identical; <sup>d</sup> if both directions use breadth-first search.

# Homework Readings

- Modern Approach Chapter 3 (3.4.2, 3.4.4, 3.4.5)
- G F Lugar Chapter 3 (3.2.4)