

```
search_bstf([Goal|Rest],Goal):-
    goal(Goal).
search_bstf([Current|Rest],Goal):-
    children(Current,Children),
    add_bstf(Children,Rest,NewAgenda),
    search_bstf(NewAgenda,Goal).

% add_bstf(A,B,C) <- C contains the elements of A and B
%                               (B and C sorted according to eval/2)
add_bstf([],Agenda,Agenda).
add_bstf([Child|Children],OldAgenda,NewAgenda):-
    add_one(Child,OldAgenda,TmpAgenda),
    add_bstf(Children,TmpAgenda,NewAgenda).

% add_one(S,A,B) <- B is A with S inserted acc. to eval/2
add_one(Child,OldAgenda,NewAgenda):-
    eval(Child,Value),
    add_one(Value,Child,OldAgenda,NewAgenda).
```

Best-first search

```
% tiles_a(A,M,V0,V) <- goal position can be reached from
%           one of the positions on A with last
%           move M (best-first strategy)
tiles_a([v(V,LastMove)|Rest],LastMove,Visited,Visited):-
    goal(LastMove).
tiles_a([v(V,LastMove)|Rest],Goal,Visited0,Visited):-
    show_move(LastMove,V),
    setof0(v(Value,NextMove),
           ( move(LastMove,NextMove),
             eval(NextMove,Value) ),
           Children),          % Children sorted on Value
    merge(Children,Rest,NewAgenda), % best-first
    tiles_a(NewAgenda,Goal,[LastMove|Visited0],Visited).
```

outOfPlace

bLeftOfw

0	●●●□○○○	9
1	●●●●○○○	9
2	●●□○○●○○	8
4	●●●○○●□○	7
5	●●○○○●●○	7
6	●●○○○□○●	6
8	●□○○○●○○●	4
9	●●○□○○●○○	4
10	□○○●○○●○○	3
12	●○○●□○○●	2
13	○○●●○○●	1
15	○○□○○●●	0

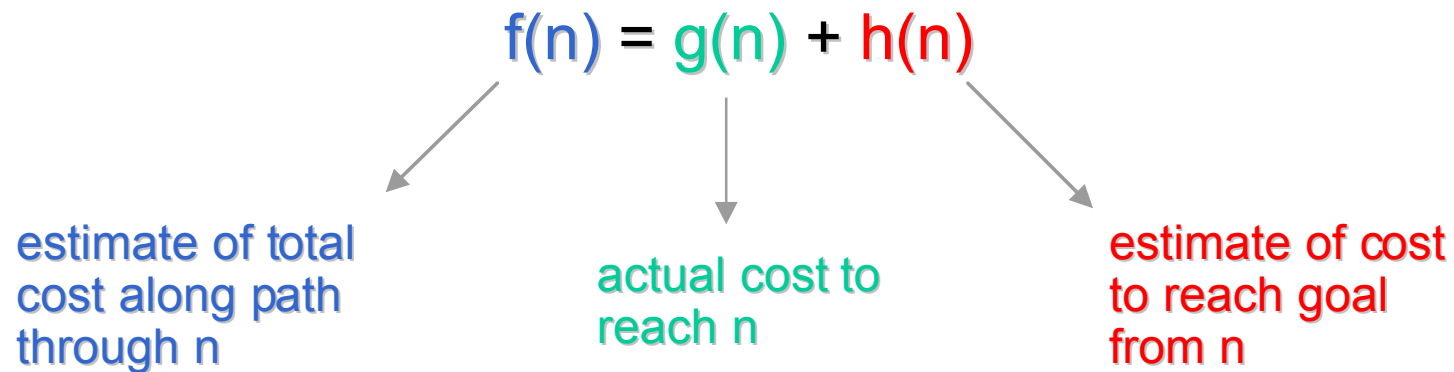


0	●●●□○○○	12
1	●●●○○○○	10
3	●●□○○●○	9
4	□●○○○●○	7
6	●●●□○○●	7
8	○○●●○○●	4
9	○●●○○○●	4
11	○●□○○●●	3
12	○○●○○●●	2
14	○○○○●●●	0



0	●●●□○○○	18
1	●●●○○○○	15
3	●●□○○●○	13
4	●●○○○●○	11
6	●□○○○●○	8
7	□●○○○●●	7
8	●●□○○●●	7
9	○□○○○●●	6
10	○○●○○●●	6
12	○○●○○●○	2
13	○○●○○●●	2
15	○○□○○●●	0

- ➔ An **A algorithm** is a best-first search algorithm that aims at minimising the **total cost** along a path from start to goal.



A algorithm

- ➡ A heuristic is (globally) **optimistic** or **admissible** if the estimated cost of **reaching a goal** is always less than the actual cost.

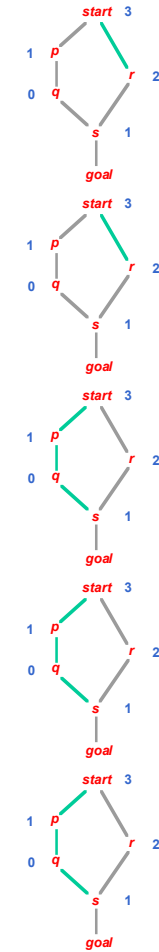
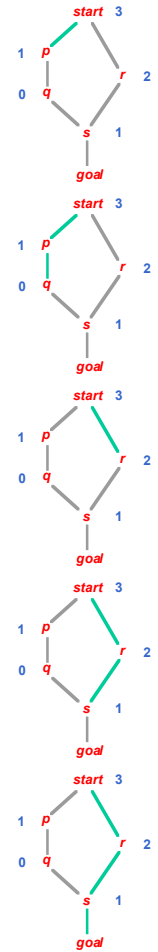
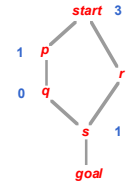
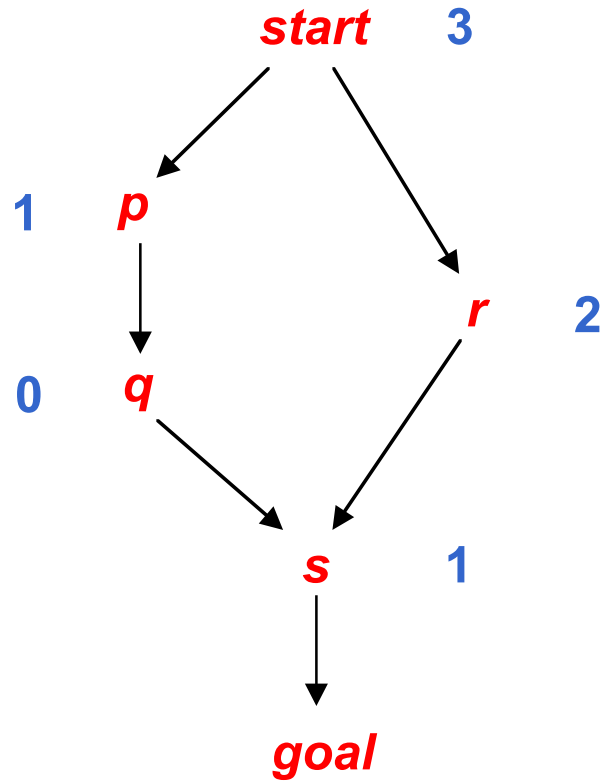
$$h(n) \leq h^*(n)$$

estimate of cost to reach goal from n

actual (unknown) cost to reach goal from n

- ➡ A heuristic is **monotonic** (locally optimistic) if the estimated cost of **reaching any node** is always less than the actual cost.

$$h(n_1) - h(n_2) \leq h^*(n_1) - h^*(n_2)$$



[start-3]

[p-2,r-3]

[q-2,r-3]

[r-3,s-4]

[s-3,s-4]

[goal-3,s-4]

Non-monotonic heuristic

```
search_beam(Agenda, Goal) :-
    search_beam(1, Agenda, [], Goal).

search_beam(D, [], NextLayer, Goal) :-
    D1 is D+1,
    search_beam(D1, NextLayer, [], Goal).
search_beam(D, [Goal | Rest], NextLayer, Goal) :-
    goal(Goal).
search_beam(D, [Current | Rest], NextLayer, Goal) :-
    children(Current, Children),
    add_beam(D, Children, NextLayer, NewNextLayer),
    search_beam(D, Rest, NewNextLayer, Goal).
```

☞ Here, the number of children to be added to the beam is made dependent on the depth **D** of the node

✓ in order to keep depth as a 'global' variable, search is layer-by-layer

```
search_hc(Goal, Goal):-
```

```
    goal(Goal).
```

```
search_hc(Current, Goal):-
```

```
    children(Current, Children),
```

```
    select_best(Children, Best),
```

```
    search_hc(Best, Goal).
```

```
% hill_climbing as a variant of best-first search
```

```
search_hc([Goal|_], Goal):-
```

```
    goal(Goal).
```

```
search_hc([Current|_], Goal):-
```

```
    children(Current, Children),
```

```
    add_bstf(Children, [], NewAgenda),
```

```
    search_hc(NewAgenda, Goal).
```