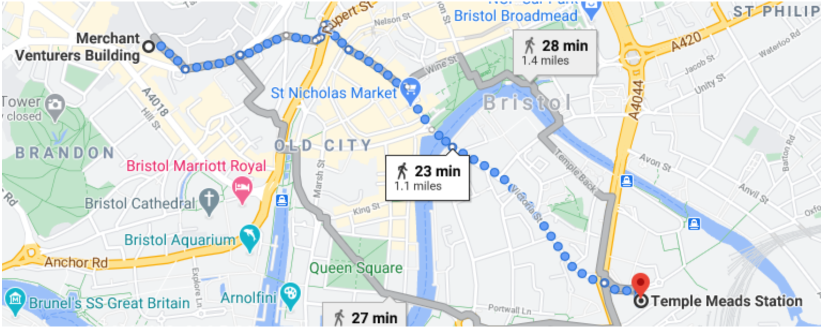
Dijkstra's algorithm
COMS20010 (Algorithms II)

John Lapinskas, University of Bristol

Distances in real networks are weighted!

We often model road networks as graphs: junctions and destinations are vertices, roads are edges, one-way roads are directed edges.

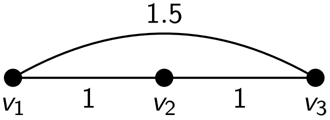


But when we want to find a "shortest path" in this graph, we don't care about the number of edges, we care about the **physical distance**.

(We may also want to weight by e.g. elevation changes or current traffic.)

Weighted graphs

A **weighted graph** is a pair (G, w), where G is a graph and $w \colon E(G) \to \mathbb{R}$ is a **weight function**. This could represent distances, costs, times, etc.



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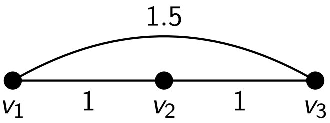

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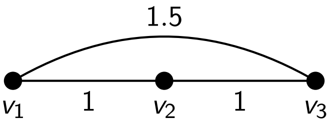

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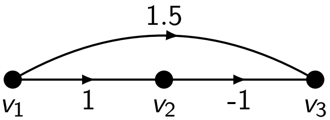
The **distance** from x to y is the shortest length of any path/walk from x to y, or ∞ if they are in different components. E.g. $d(v_1, v_3) = 1.5$.

Negative-weight edges

For some applications, it can make sense to allow edges to have **negative weight**. (E.g. costs versus profits...) This can be counterintuitive!



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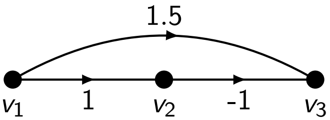

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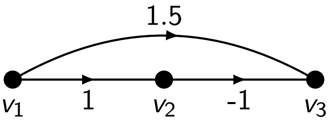

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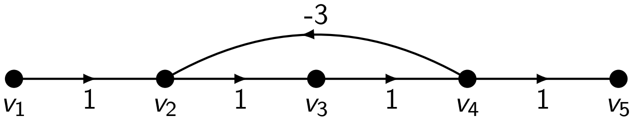

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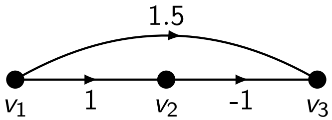

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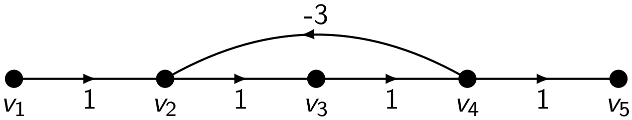

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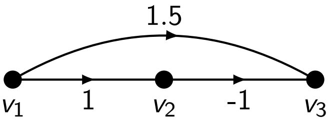

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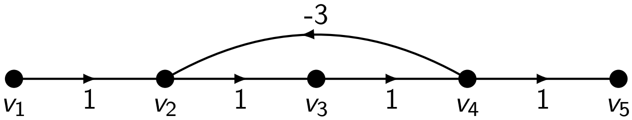

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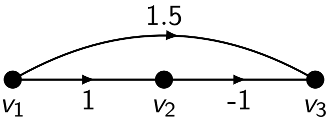

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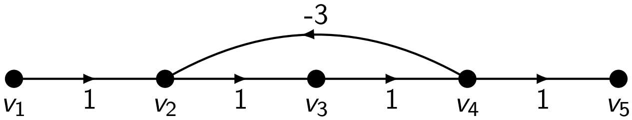

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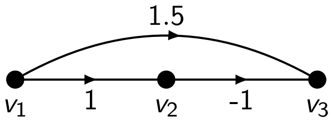

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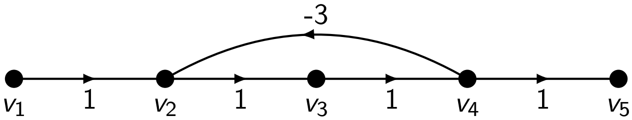

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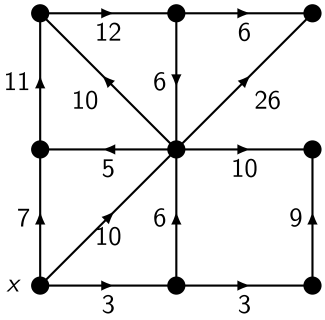
This lecture, we ignore negative weights. (This is also faster!)

Dijkstra's algorithm: The idea



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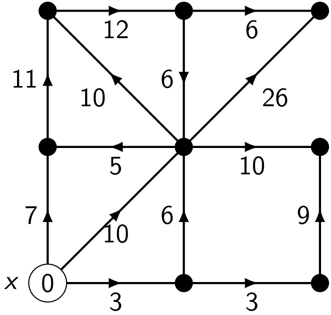
When the water first reaches a vertex v, you know $d(x, v)$ and a shortest path from x to v.



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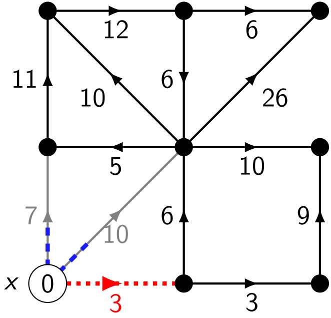

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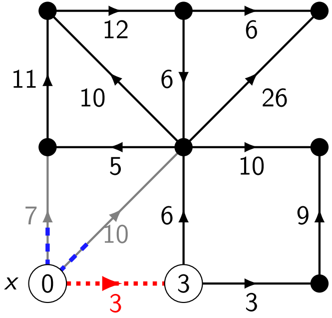

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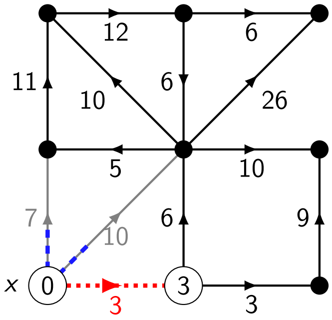

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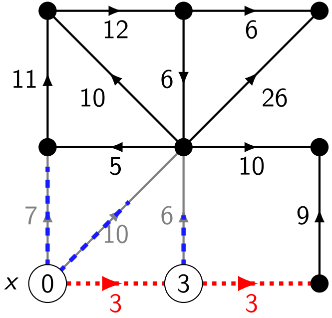

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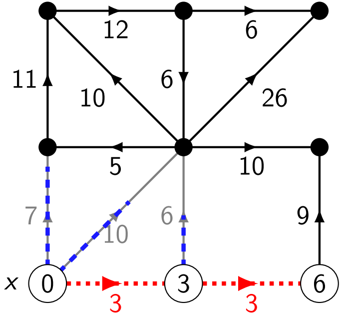

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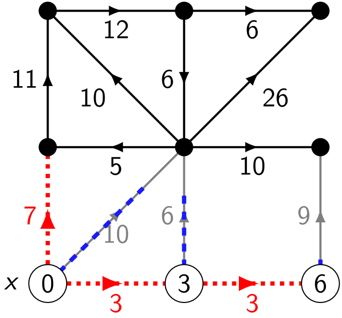

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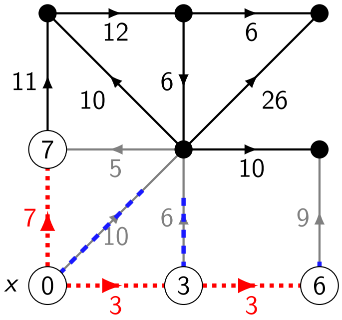

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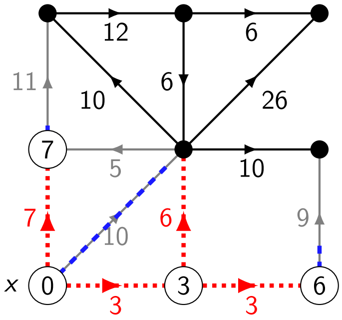

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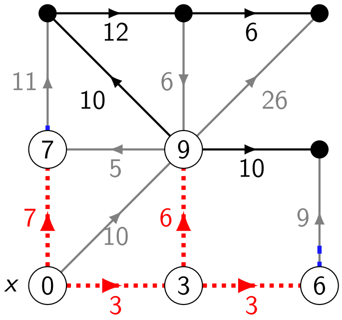

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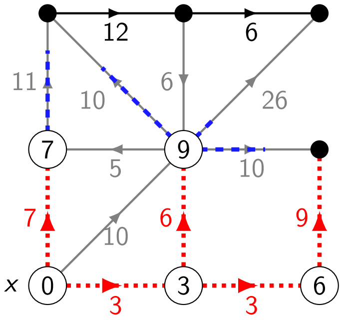

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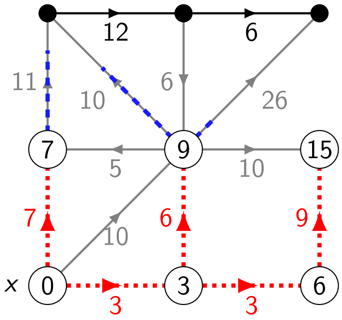

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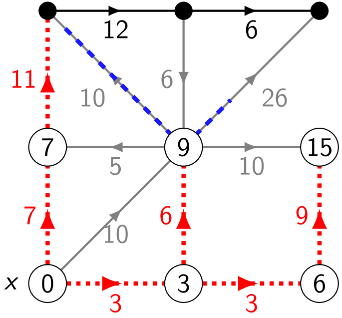

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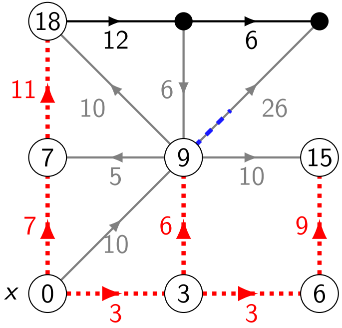

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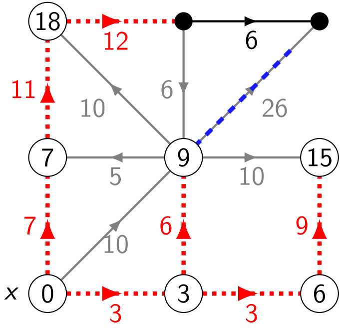

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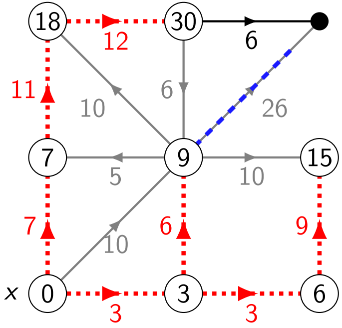

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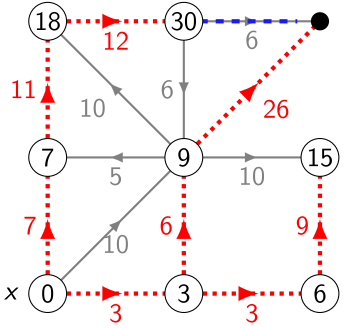

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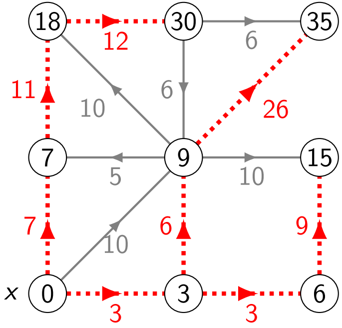

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Proof: By induction on the number of vertices u with $d(x, u)$ set.

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For input graph G, vertex x: At each stage, pick an edge (u, v) with $d(x, u)$ known and $d(x, v)$ unknown that minimises $d(x, u) + \mathrm{length}(u, v)$. Then set $d(x, v) = d(x, u) + \mathrm{length}(u, v)$.

Claim: Dijkstra's algorithm calculates distances correctly.

Proof: By induction on the number of vertices u with $d(x, u)$ set.

Base case: We have $d(x, x) = 0$. ✓

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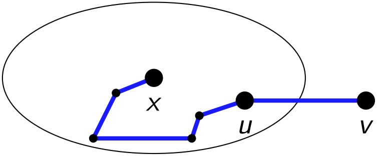

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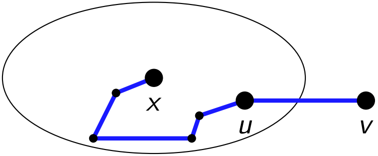

We can append (u, v) to any path from x to u, so we have

$$d(x, v) \leq d(x, u) + \mathrm{length}(u, v).$$

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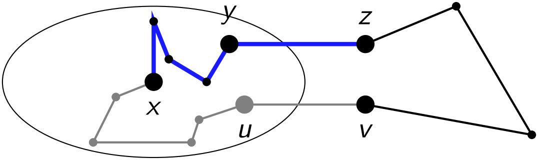

Also, any path P from x to v has to leave X on some edge (y, z).

Dijkstra's algorithm: Correctness

For input graph G, vertex x: At each stage, pick an edge (u, v) with $d(x, u)$ known and $d(x, v)$ unknown that minimises $d(x, u) + \mathrm{length}(u, v)$. Then set $d(x, v) = d(x, u) + \mathrm{length}(u, v)$.

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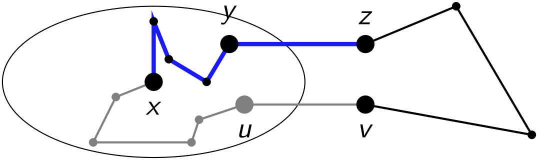

Also, any path P from x to v has to leave X on some edge (y, z). Hence P has length at least $d(x, y) + \mathrm{length}(y, z)$. So from the way we picked (u, v), we have $d(x, v) \geq d(x, u) + \mathrm{length}(u, v)$. □

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We need a **priority queue** (see COMS10007) to implement this efficiently.

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```
StartQueue(5);
```

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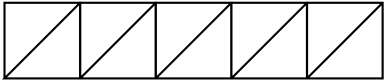

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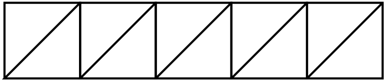

`StartQueue`(5); `Insert`$(v_1, 0)$;

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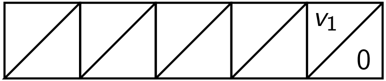

`StartQueue`(5); `Insert`$(v_1, 0)$;

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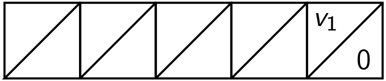

`StartQueue`(5); `Insert`$(v_1, 0)$; `Insert`$(v_2, 4)$;

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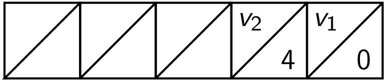

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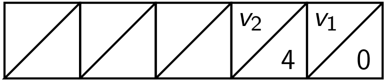

`Insert`$(v_3, 2)$;

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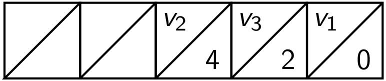

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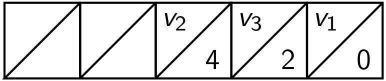

`Insert`$(v_3, 2)$; `Insert`$(v_5, 3)$;

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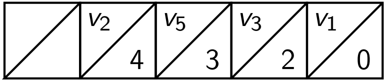

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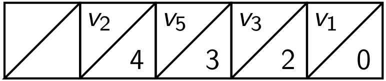

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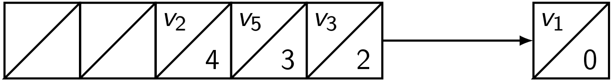

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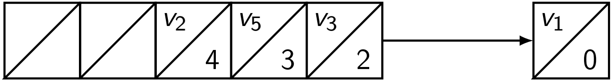

`Insert`(v_4, 20);

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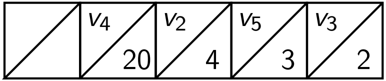

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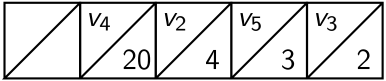

`Insert`(v_4, 20); `ChangeKey`(v_4, 1);

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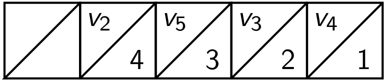

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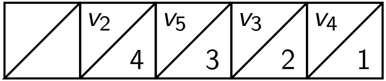

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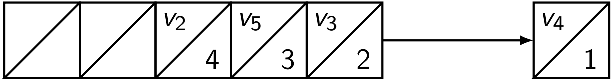


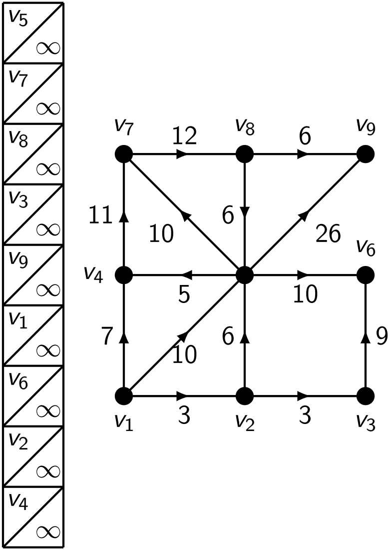
$\texttt{Insert}(v_4, 20); \texttt{ChangeKey}(v_4, 1); \texttt{Extract}();$

Dijkstra's algorithm: Implementation

Algorithm: DIJKSTRA

Input : Weighted graph $G = ((V, E), w)$, $v \in V$.
Output : $d(v, y)$ for all $y \in V$.

Number the vertices of G as $v = v_1, \dots, v_n$.
`queue` $\leftarrow$ `StartQueue`(n).
foreach $i = 1$ *to* n **do**
    `dist`$[i] \leftarrow \infty$ and call `queue.Insert`(v_i, ∞).
Call `queue.ChangeKey`$(v_1, 0)$.
do
    $v_i \leftarrow$ `queue.Extract`$()$.
    foreach $(v_i, v_j) \in E$ **do**
        `dist`$[j] \leftarrow \min\{$`dist`$[j],$ `dist`$[i] + w(i, j)\}$.
        Call `queue.ChangeKey`$(v_j,$ `dist`$[j])$,
while `queue` *is not empty*
Return `dist`.



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```
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foreach i = 1 to n do
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Call queue.ChangeKey(v_1, 0).
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    foreach (v_i, v_j) ∈ E do
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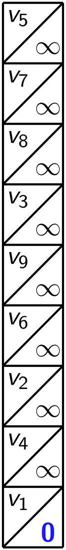

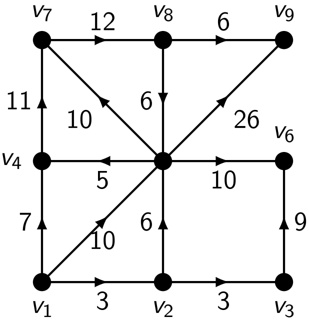

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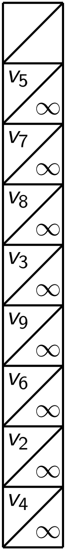

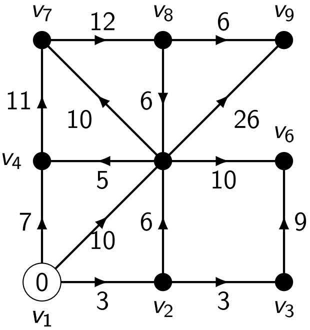

Dijkstra's algorithm: Implementation

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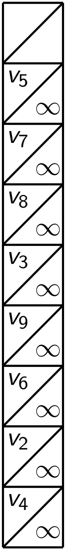

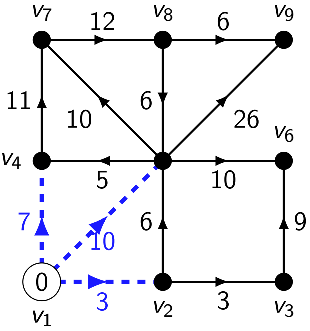

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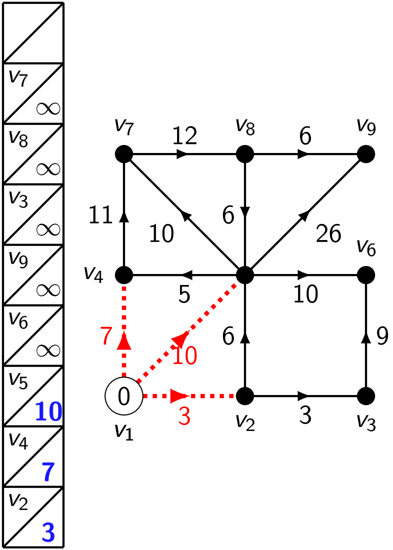

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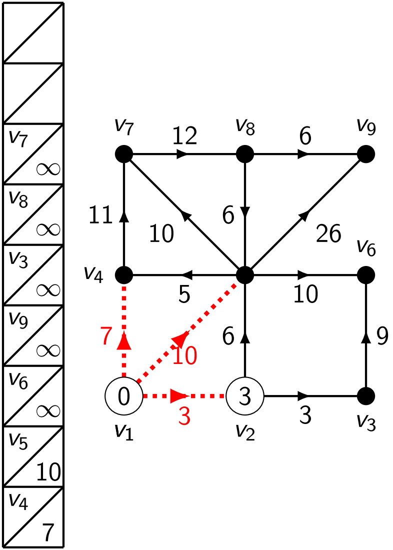

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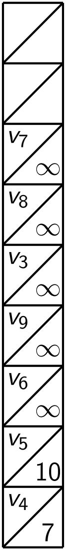

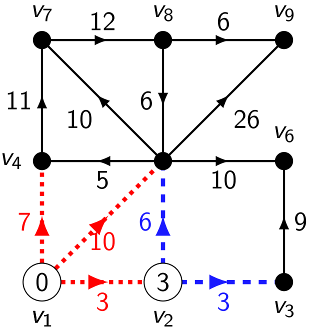

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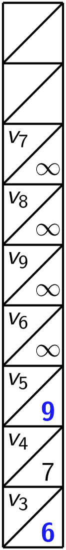

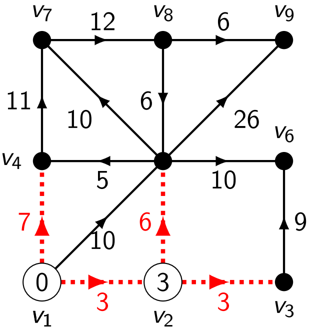

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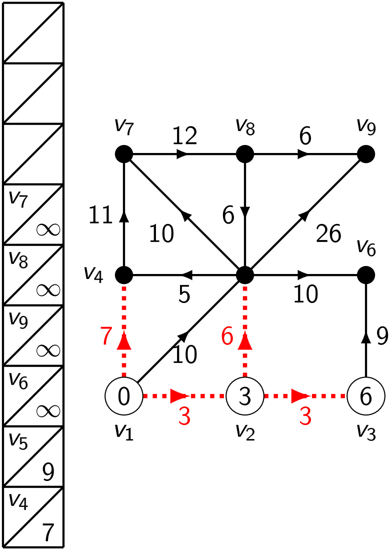

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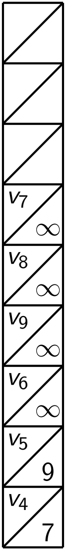

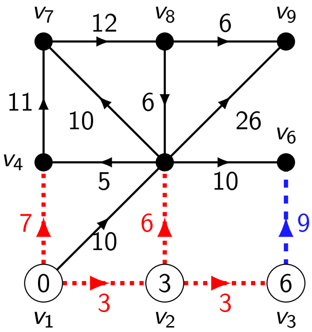

Dijkstra's algorithm: Implementation

Algorithm: Dijkstra

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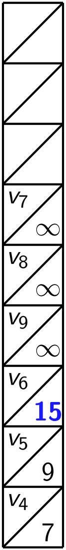

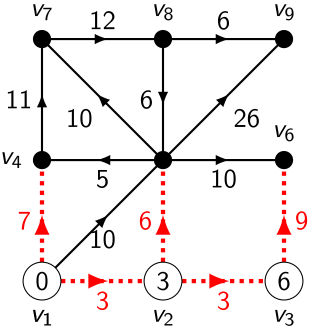

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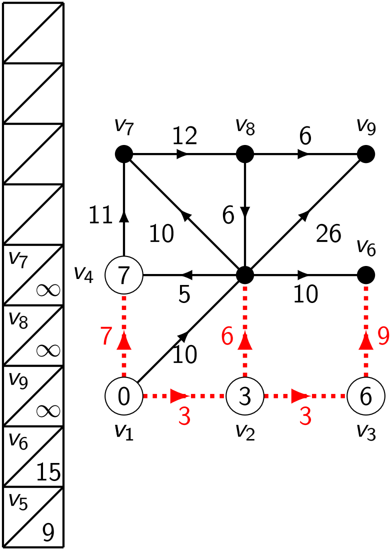

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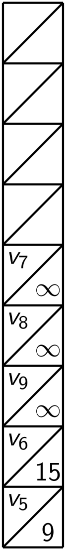

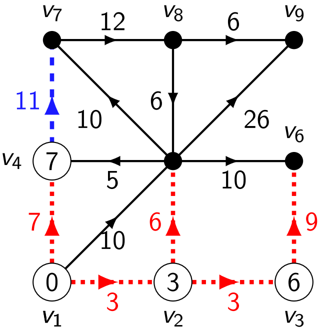

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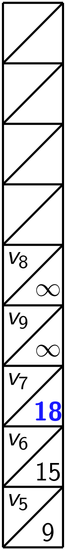

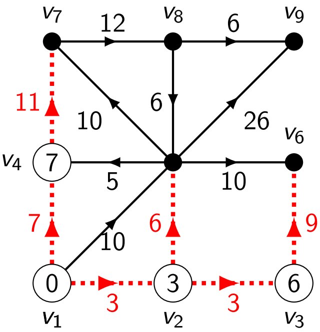

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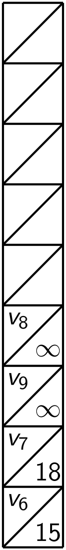

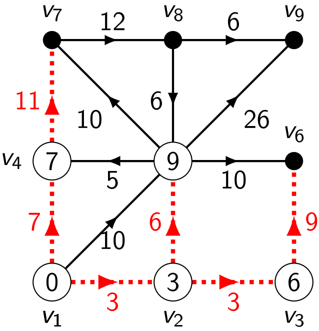

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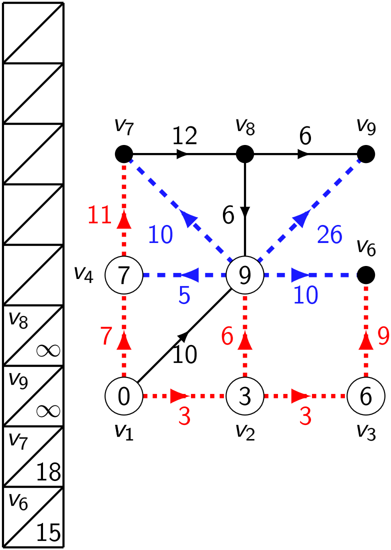

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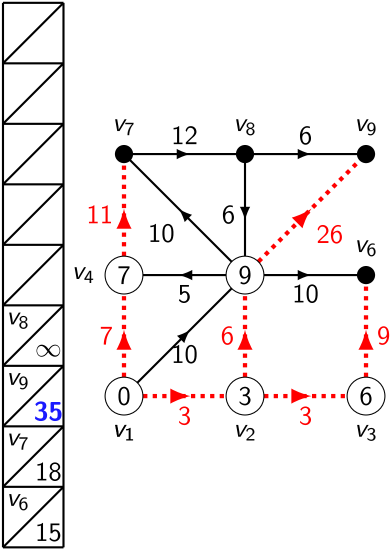

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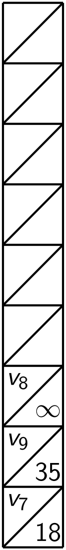

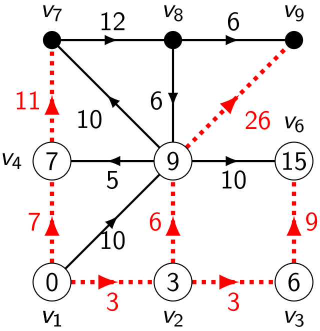

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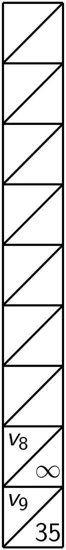

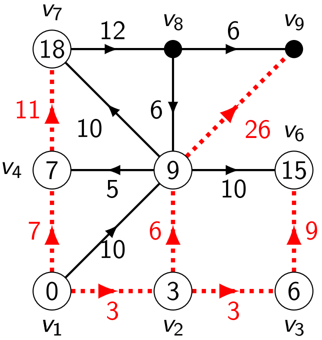

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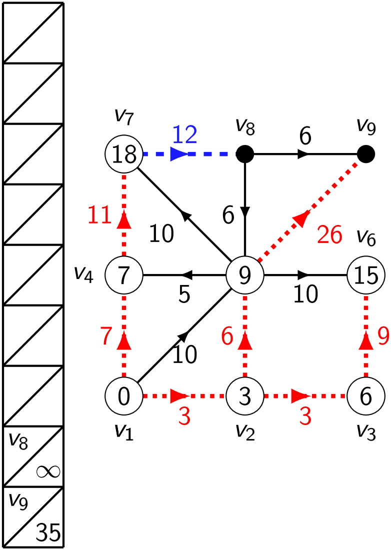

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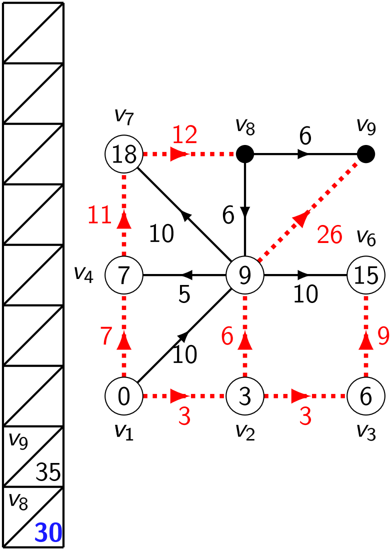

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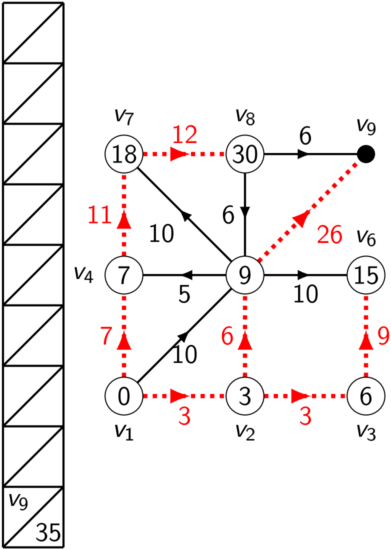

Dijkstra's algorithm: Implementation

Algorithm: DIJKSTRA

Input : Weighted graph $G = ((V, E), w)$, $v \in V$.
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Number the vertices of G as $v = v_1, \dots, v_n$.
queue $\leftarrow$ `StartQueue`(n).
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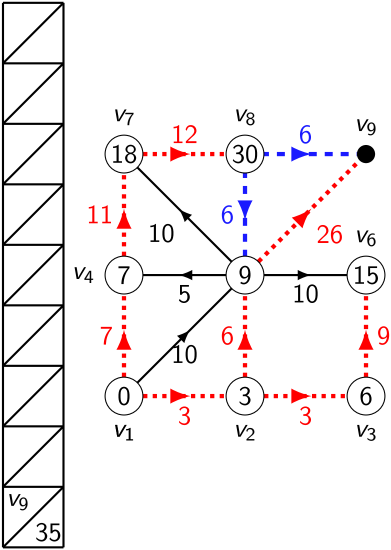

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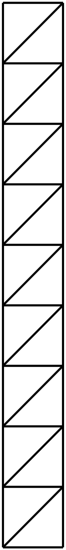

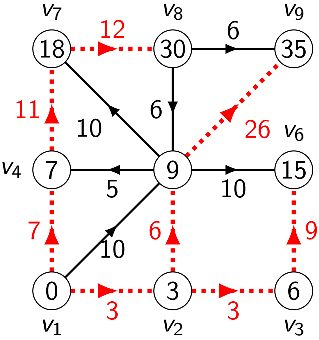

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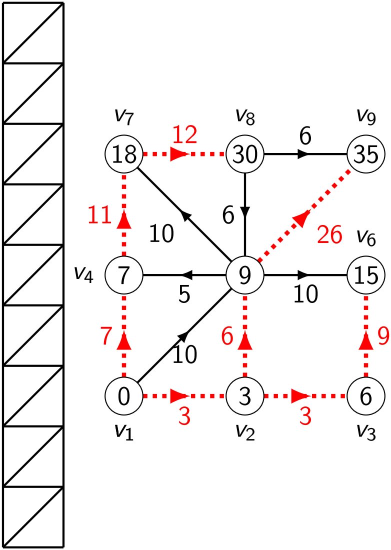
Invariant: `dist`$[j]$ is the minimum value of $d(v_1, v_i) + w(v_i, v_j)$ over all v_i's whose distances are finalised, as in mathematical version. ✓



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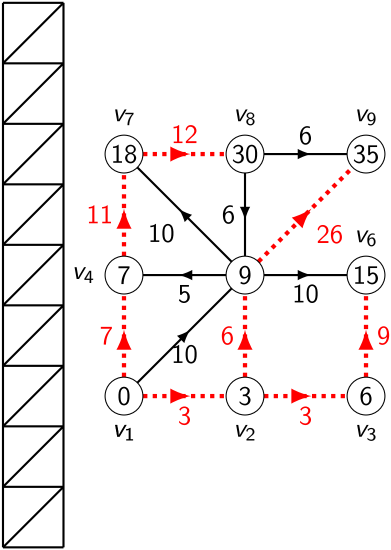
We can recover shortest paths by storing and returning the dotted red edges.



Dijkstra's algorithm: Time analysis

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We perform $O(|V|)$ `Insert` operations and `Extract` operations, and $O(|E|)$ `ChangeKey` operations, for a total of $O((|V| + |E|) \log |V|)$ time when G is given in adjacency list form.

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We could drop this to $O(|V| \log |V| + |E|)$ time by using a **Fibonacci heap** as a priority queue... But Fibonacci heaps have *awful* constants, and generally $\log |V| \lessapprox 50$, so let's not!