

# **Lecture 14**

## **Dijkstra's Algorithm**

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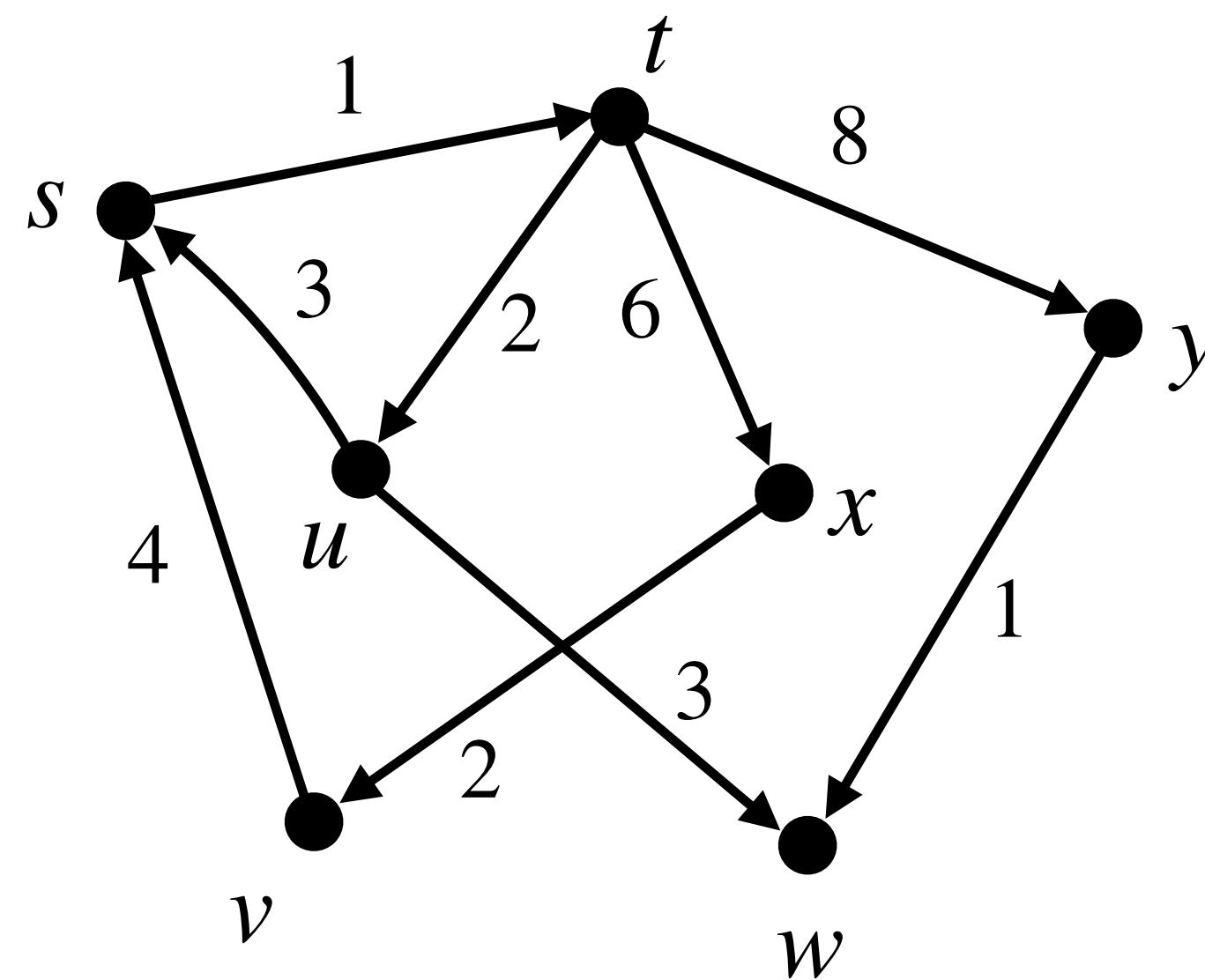
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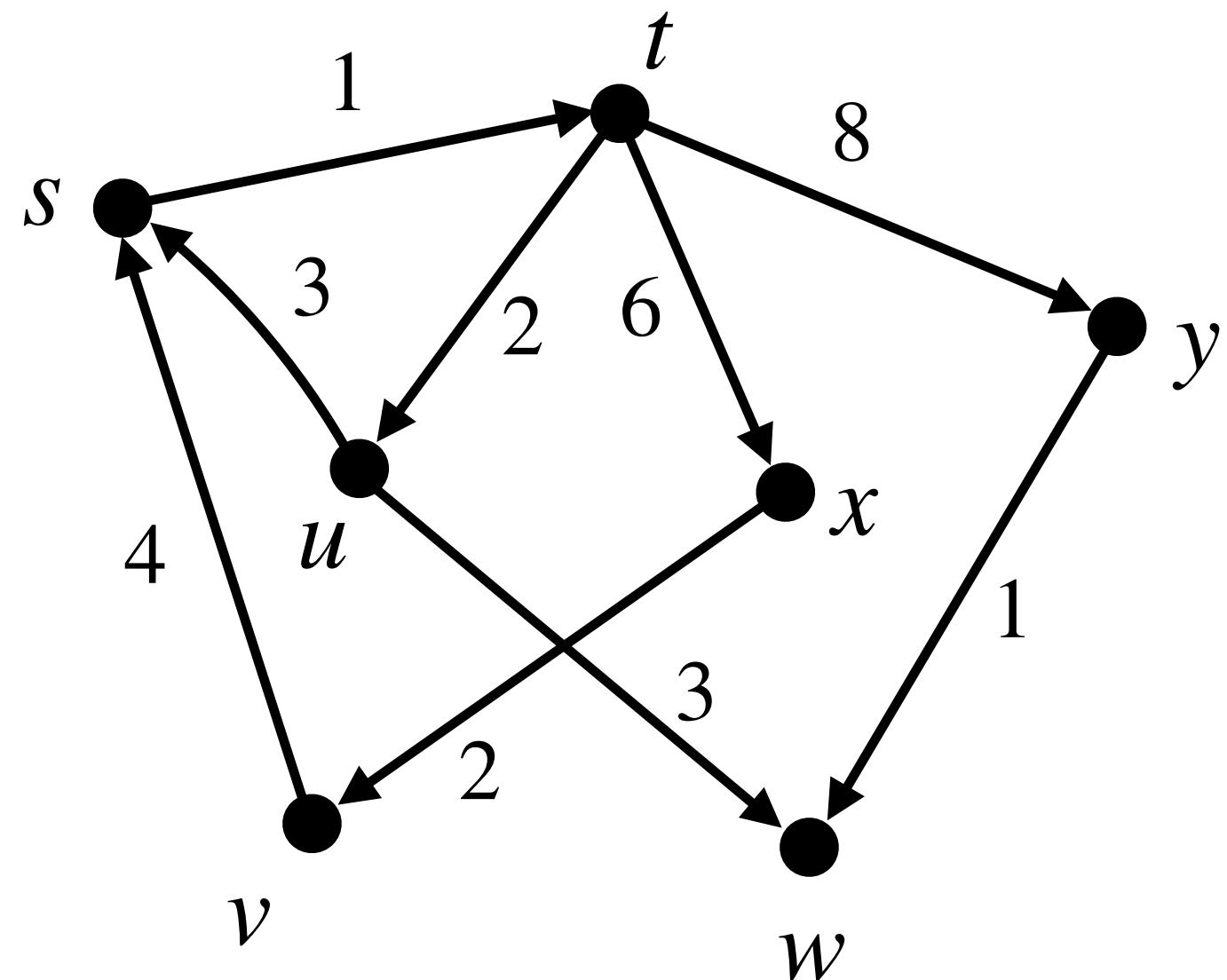
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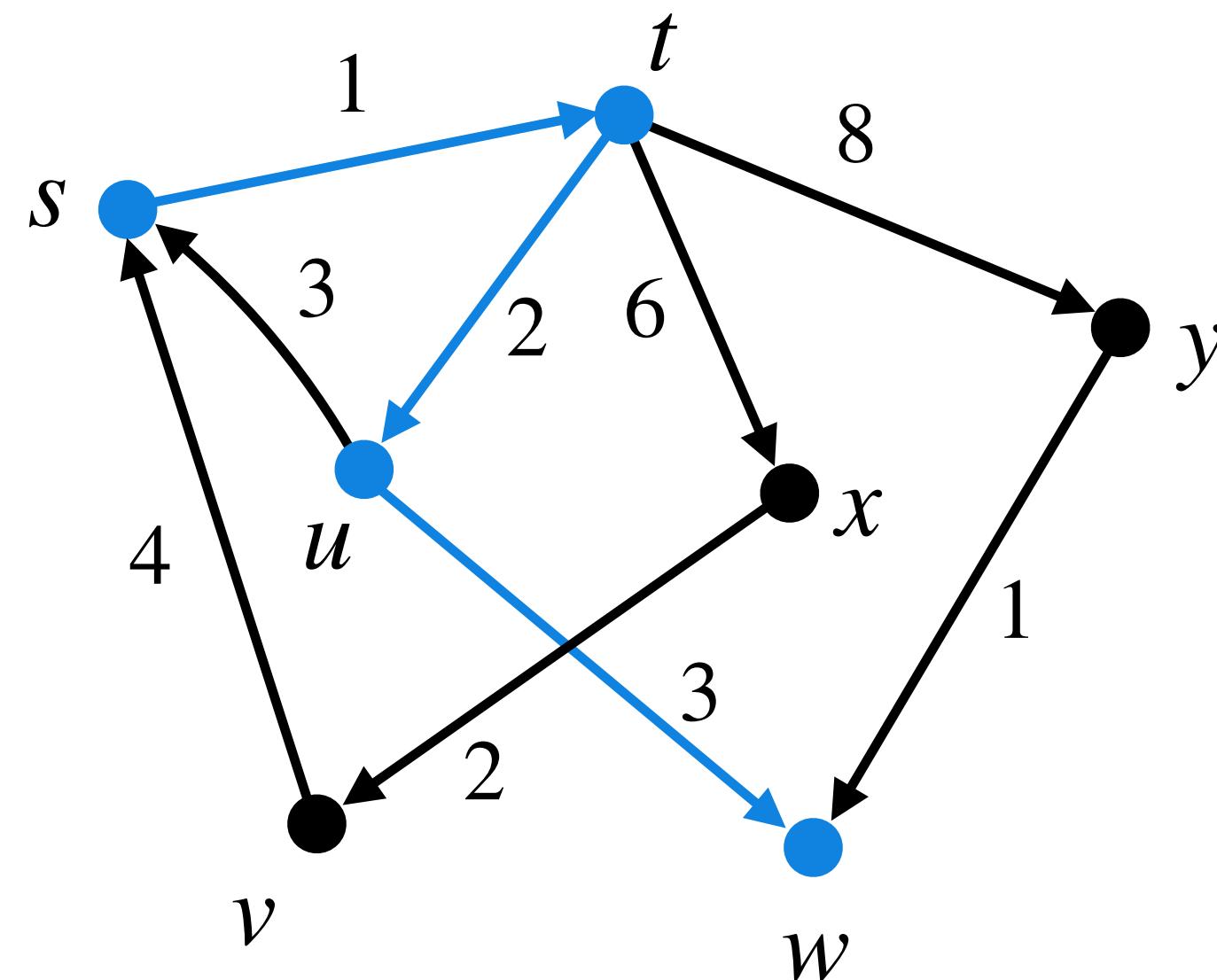
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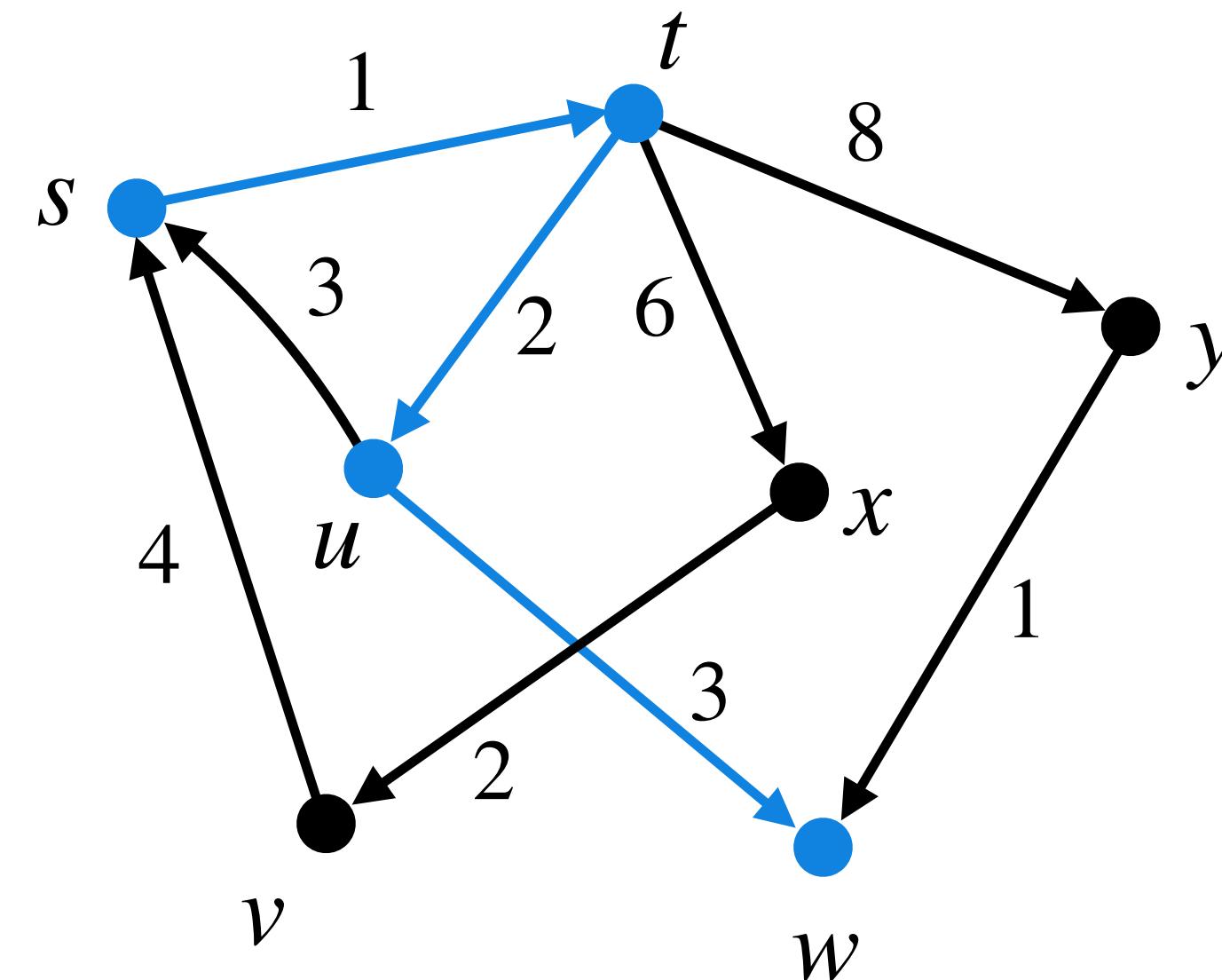
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**Example:** In the below graph shortest path from **s** to **w** is of weight 6.



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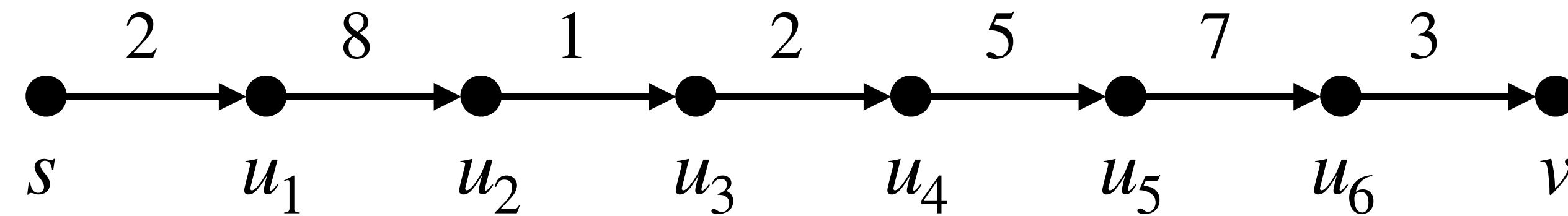
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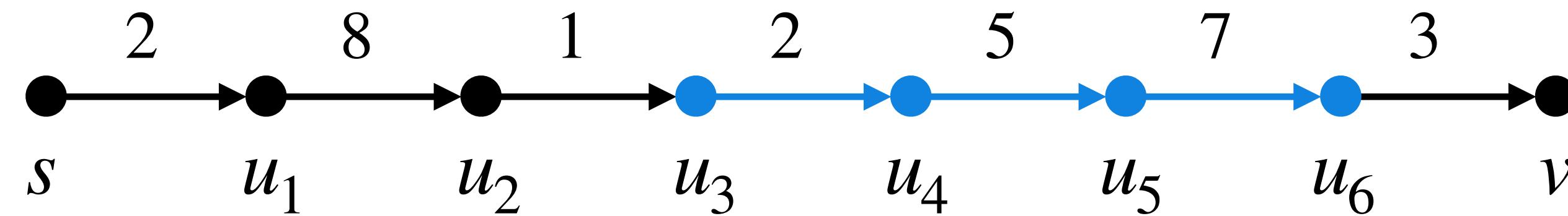
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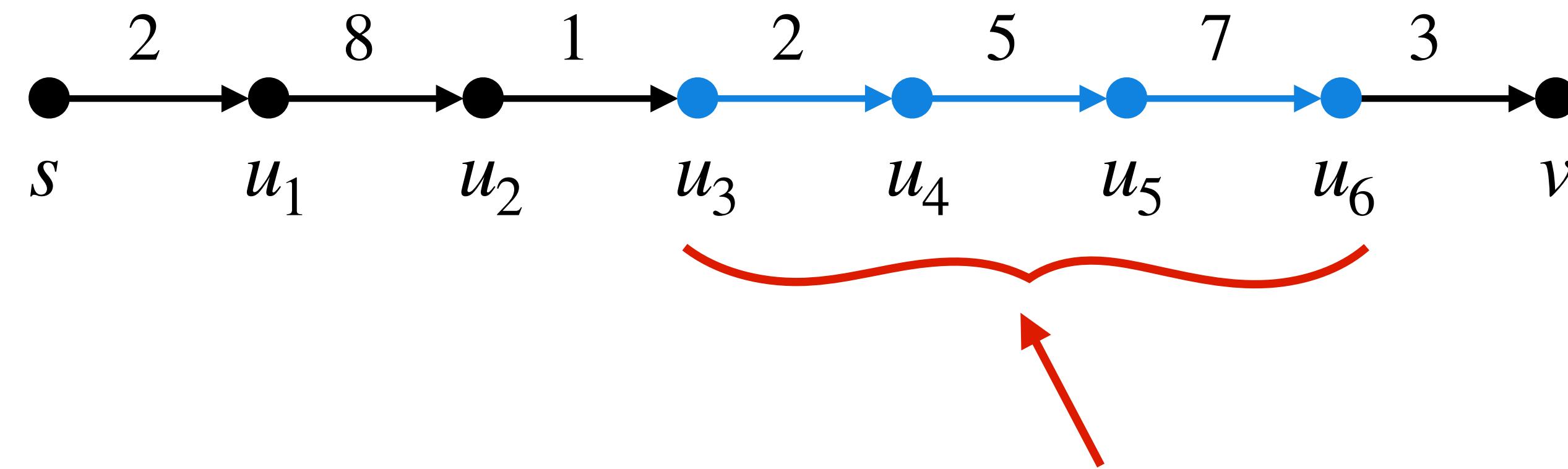
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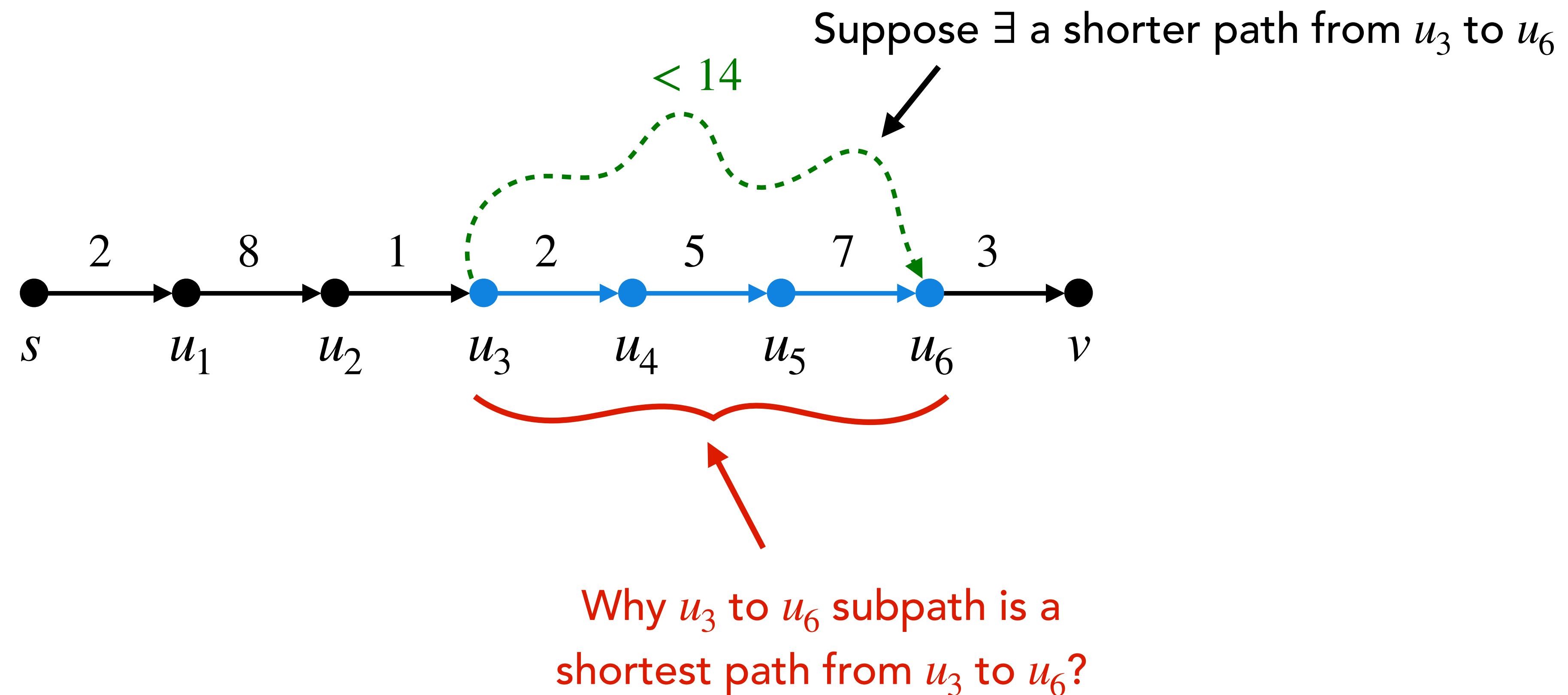


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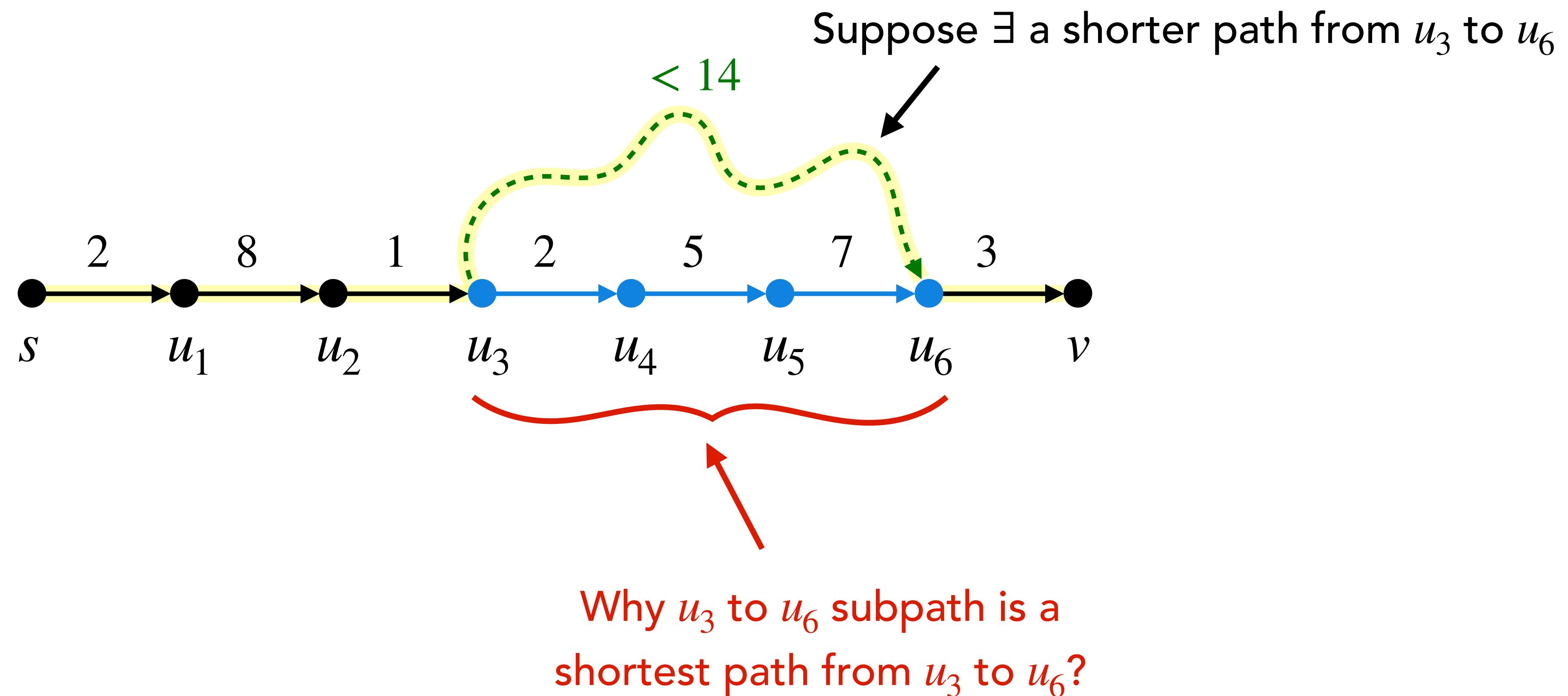
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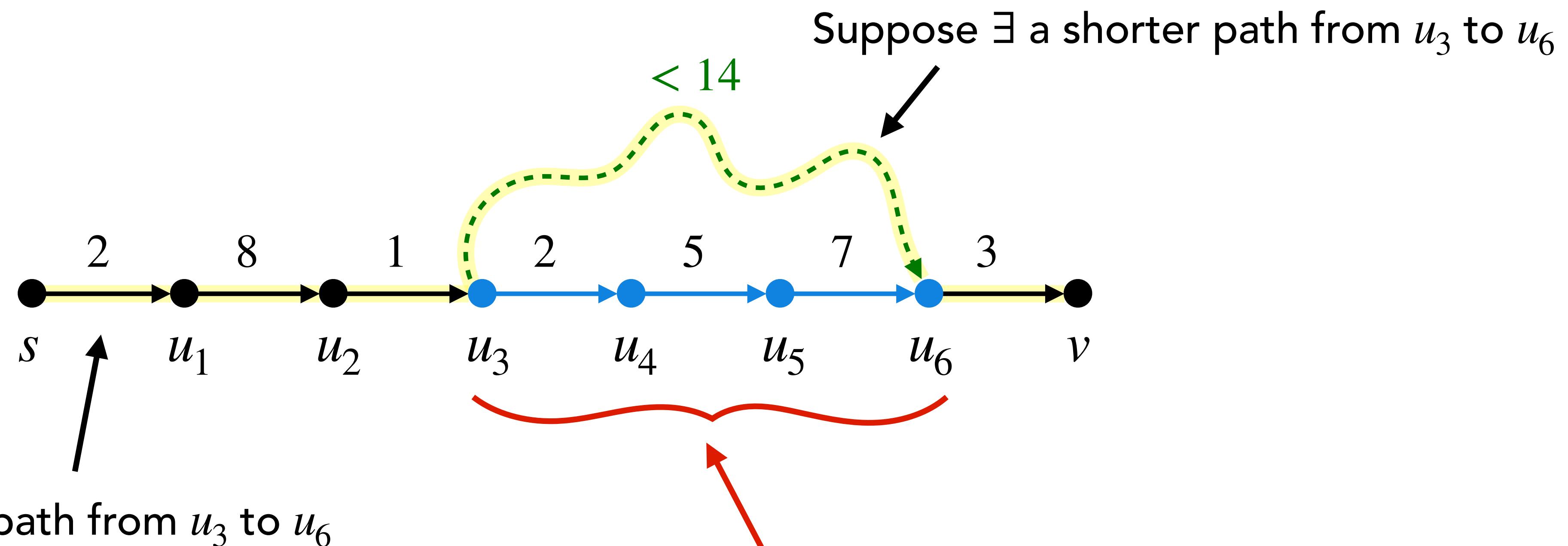
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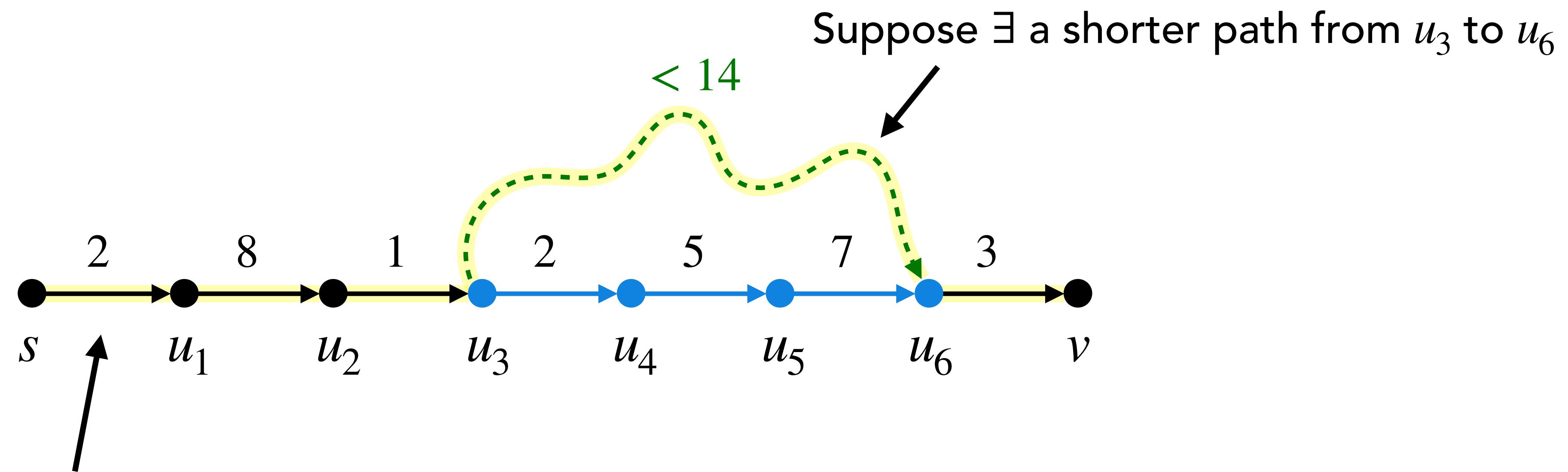


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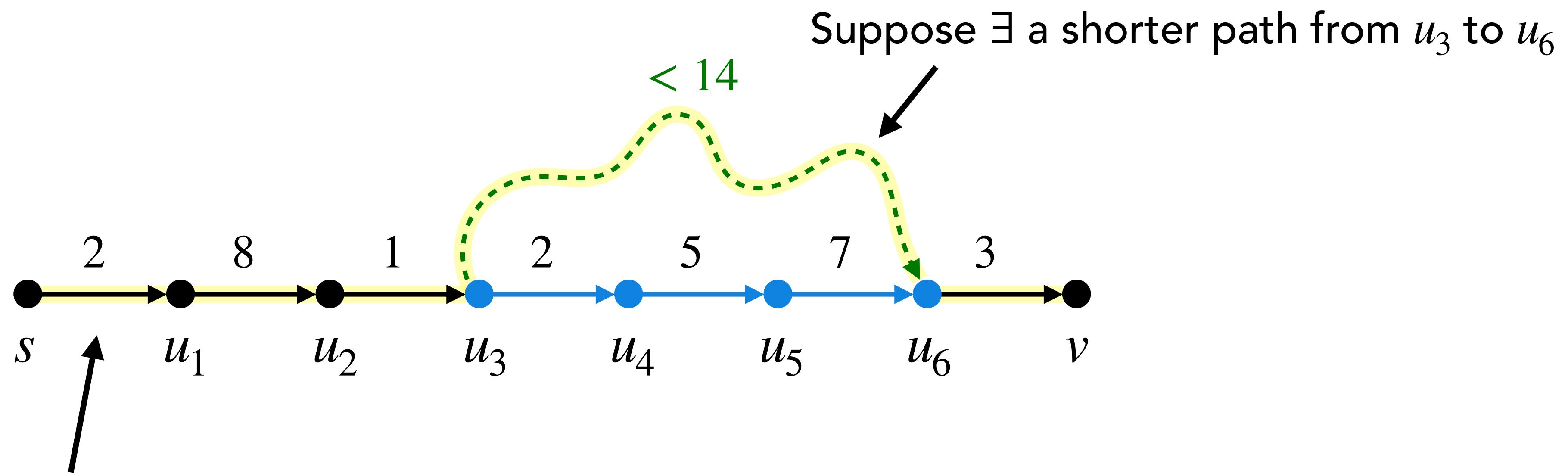
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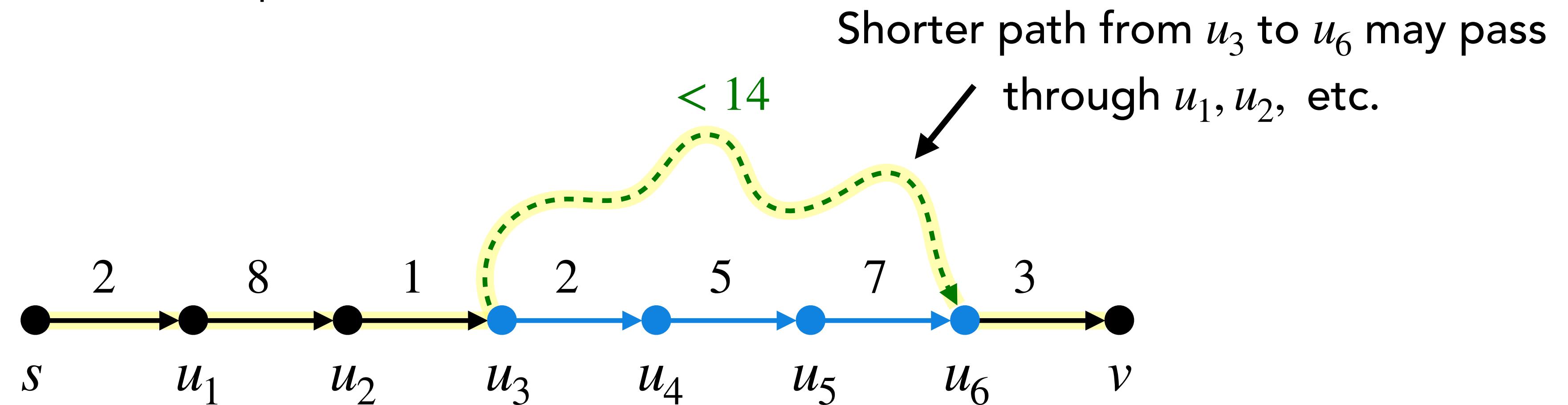
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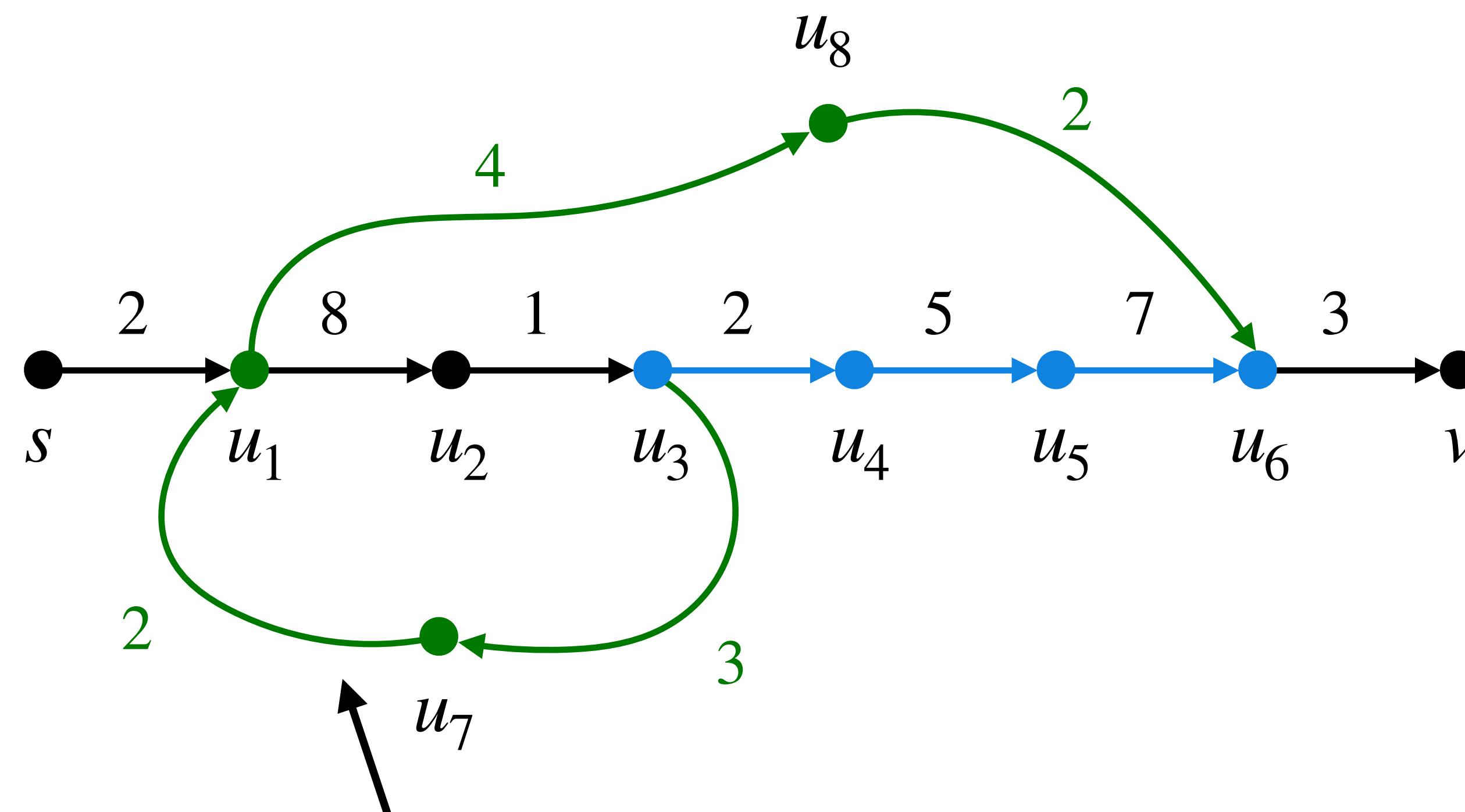
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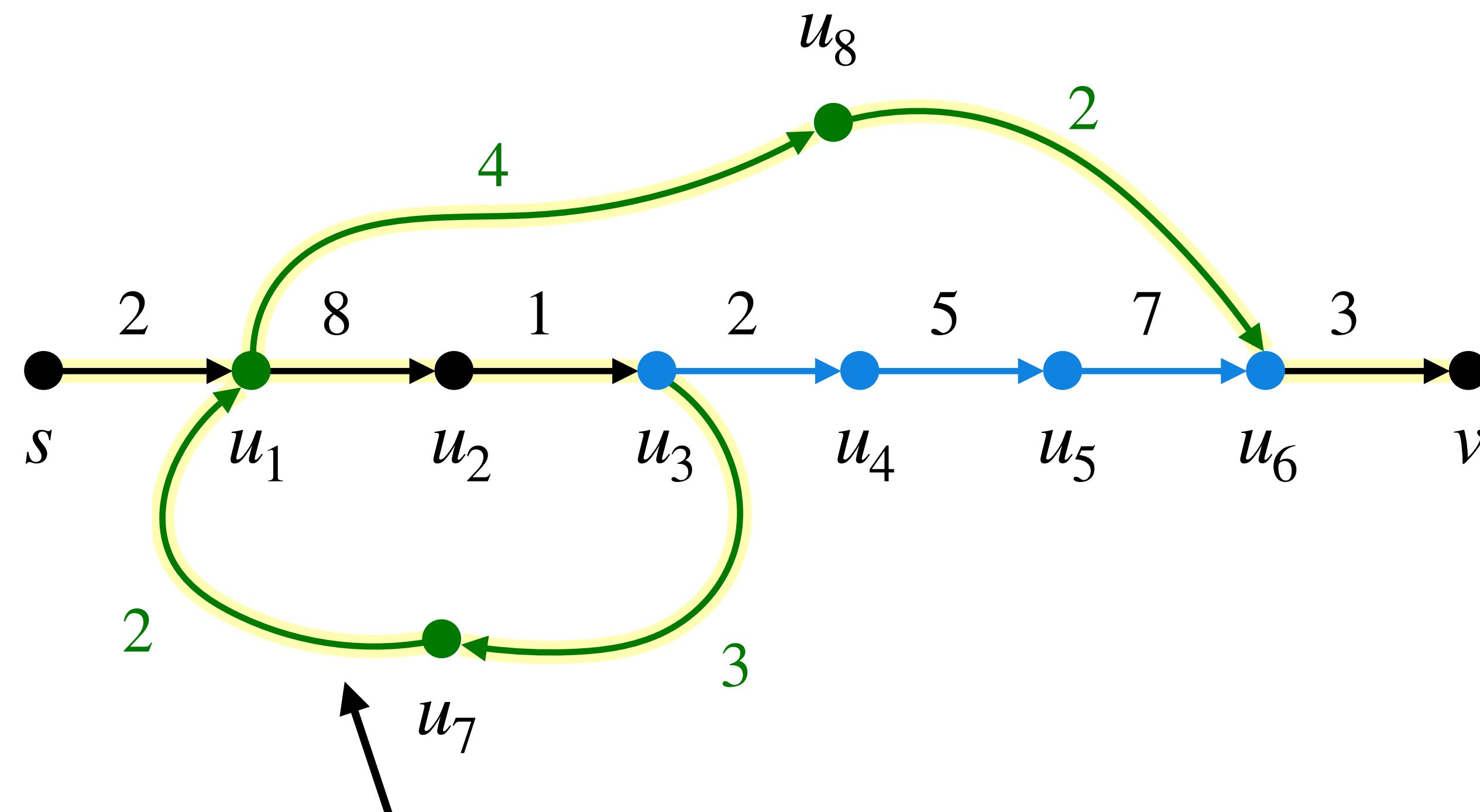


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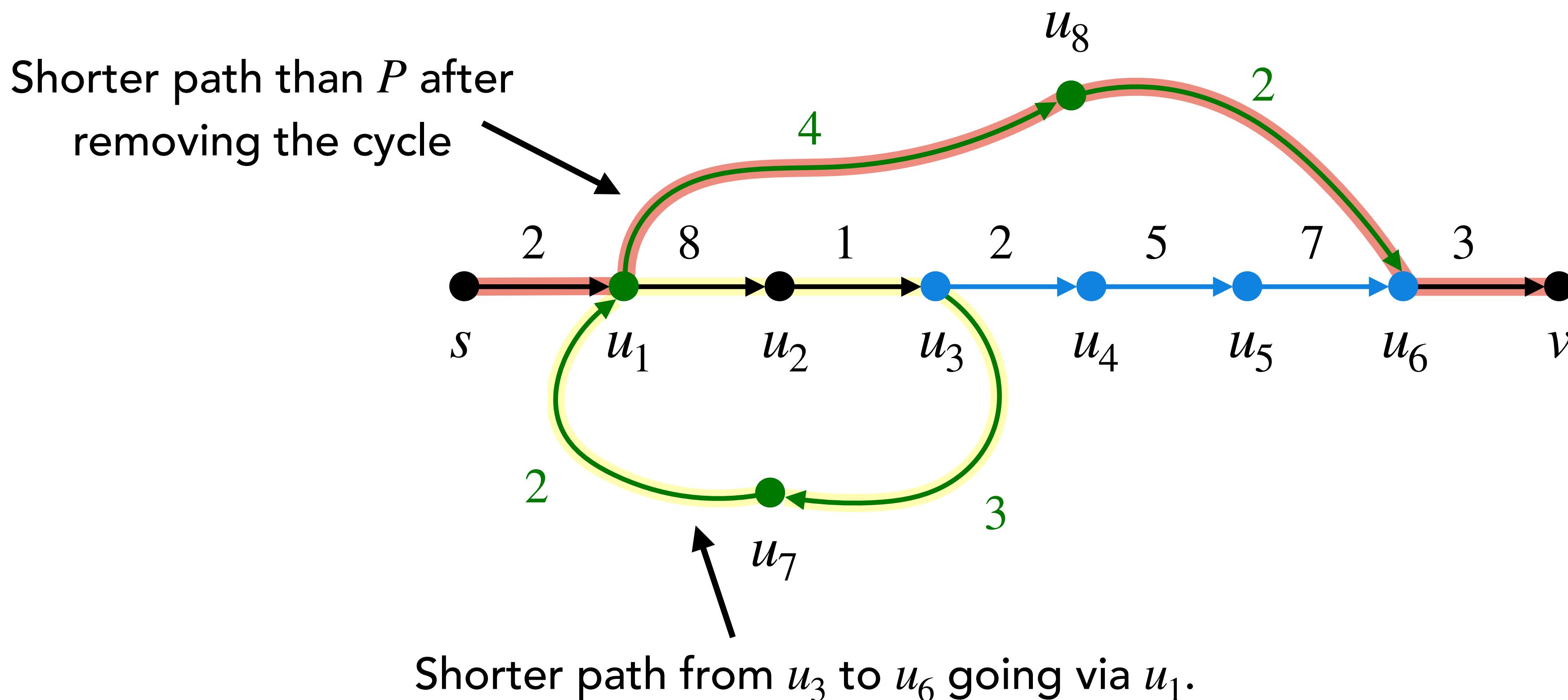


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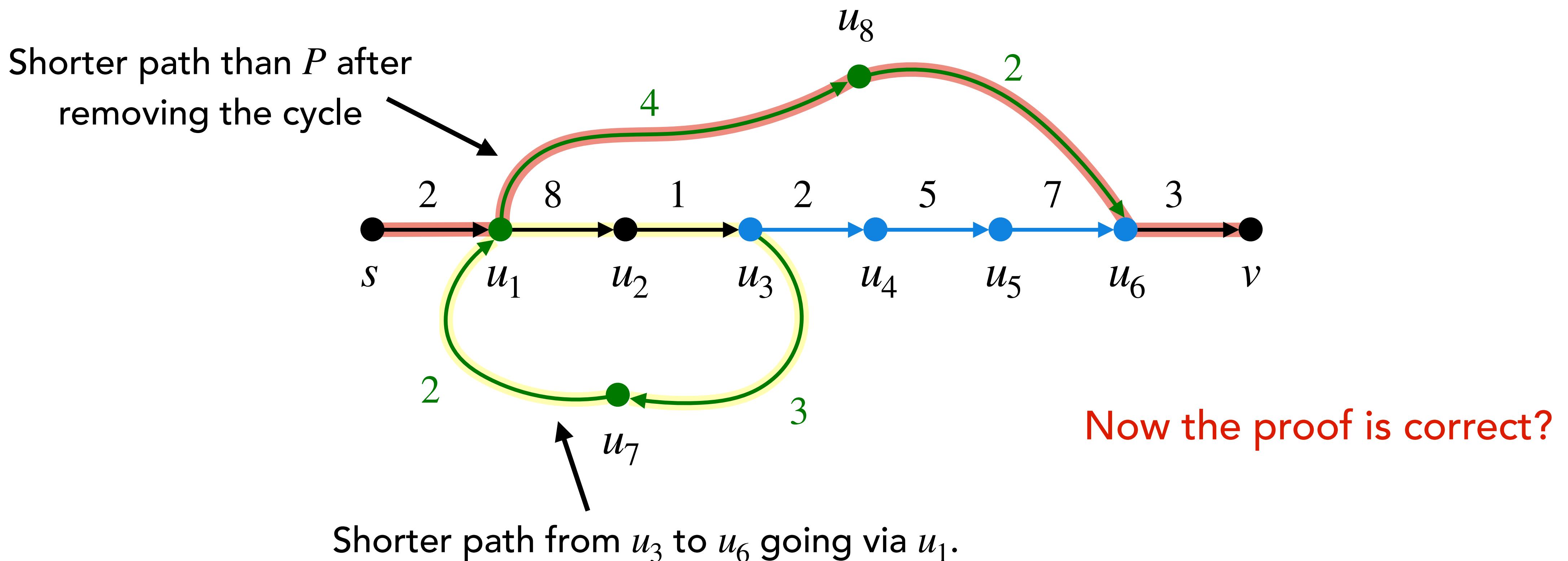
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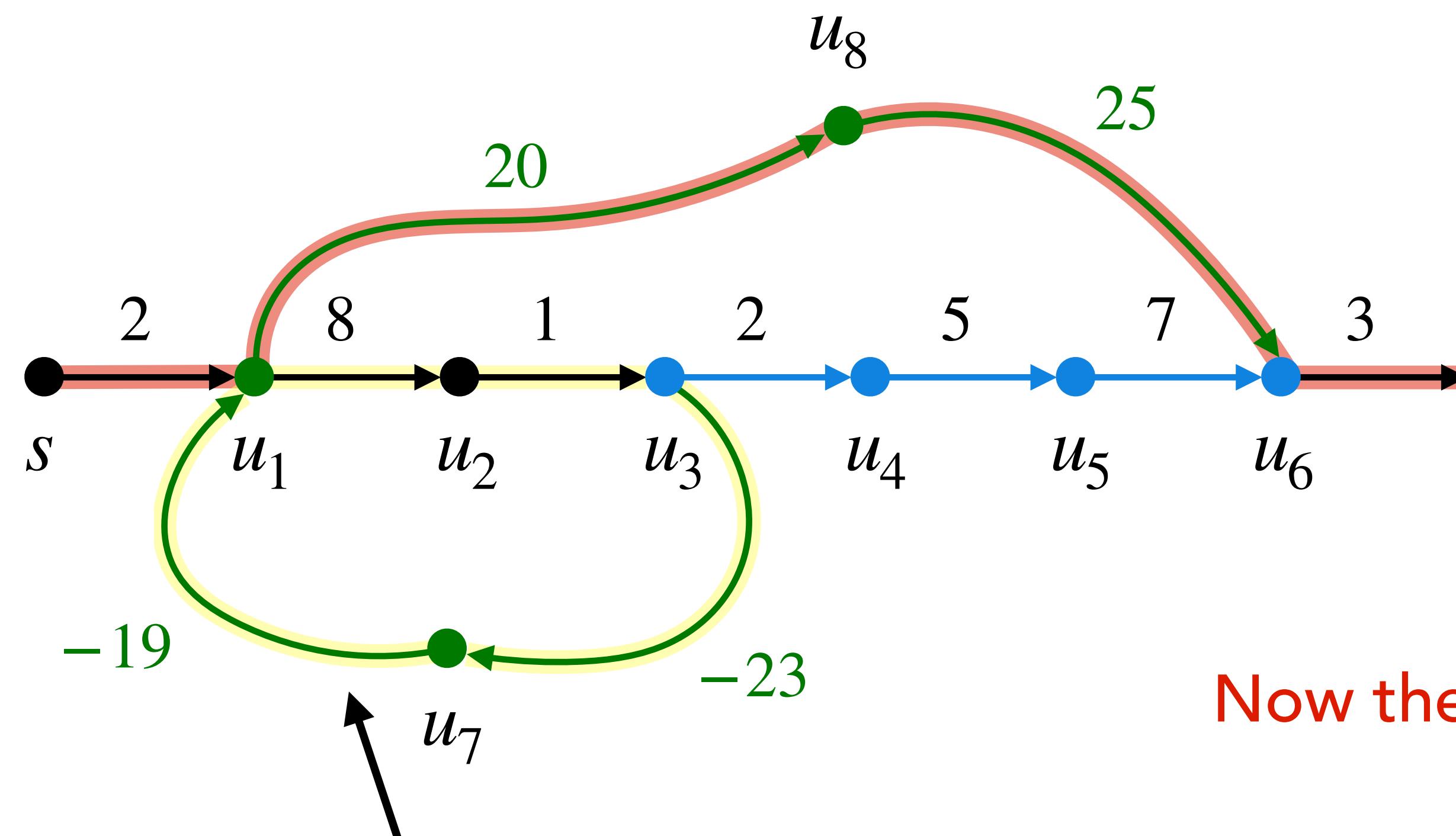
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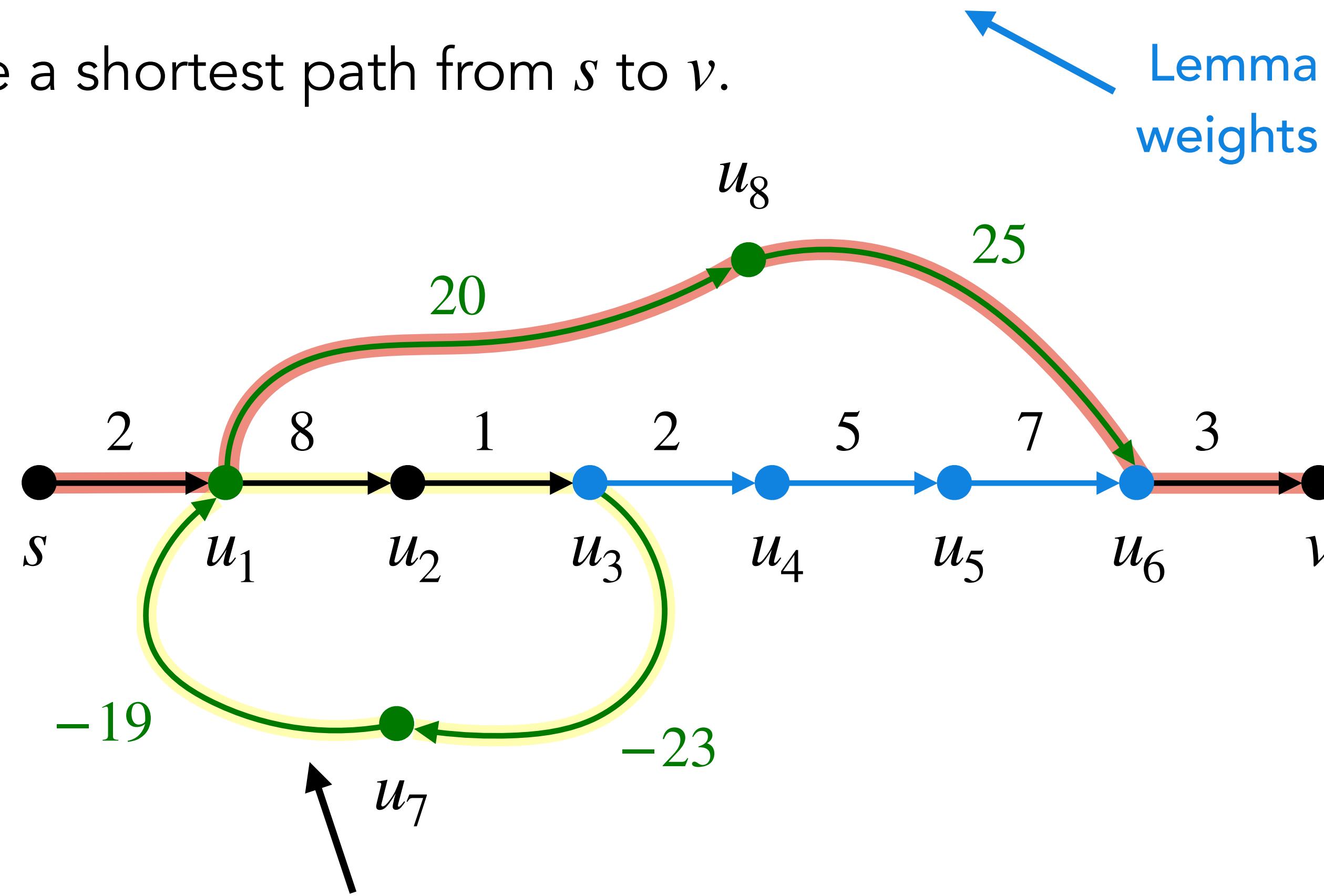
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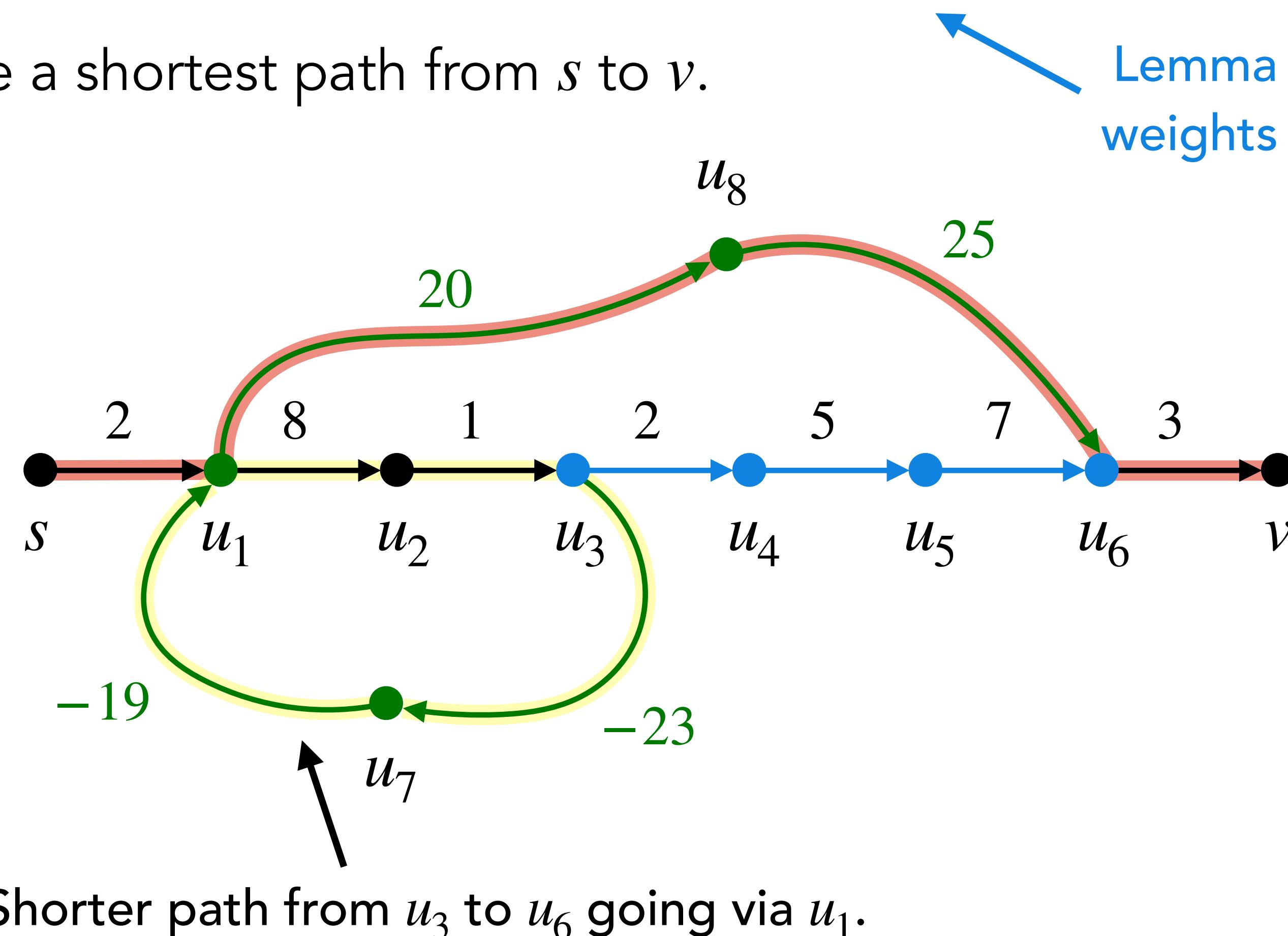
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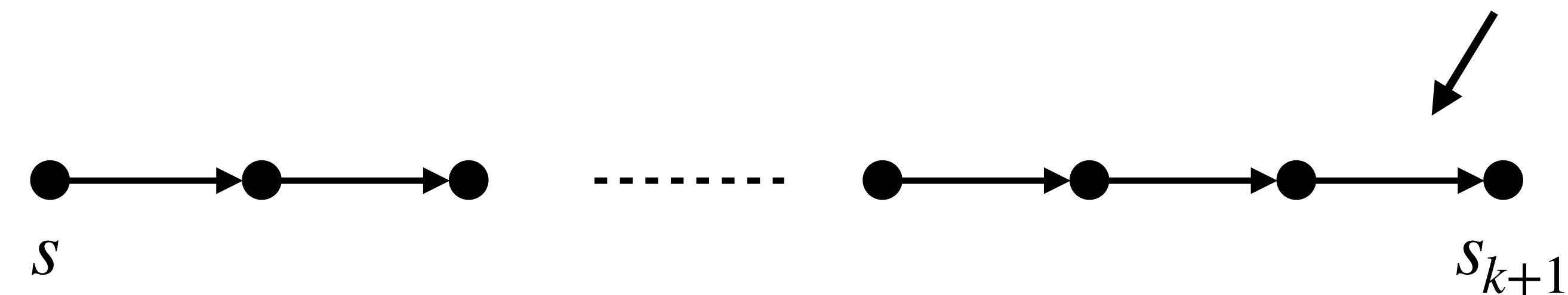
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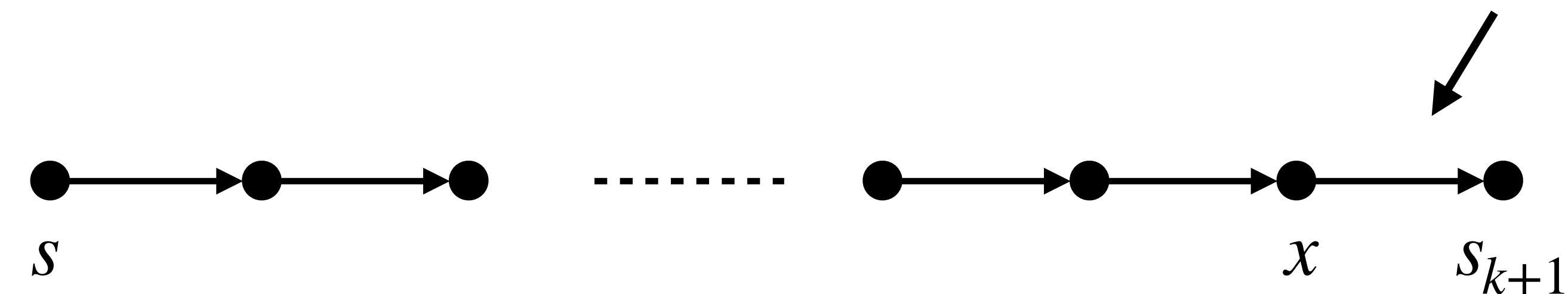
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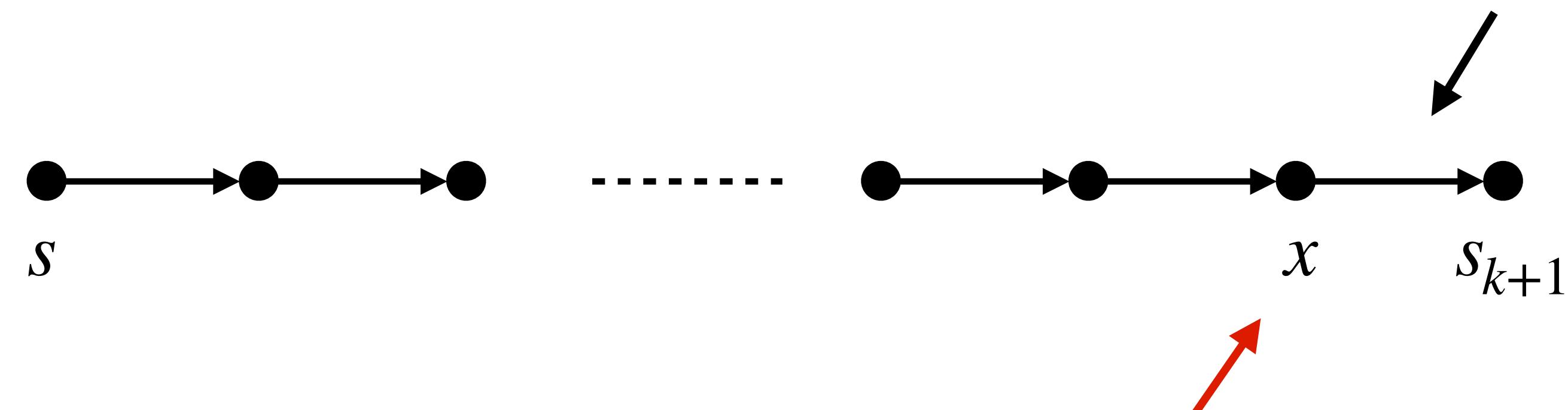
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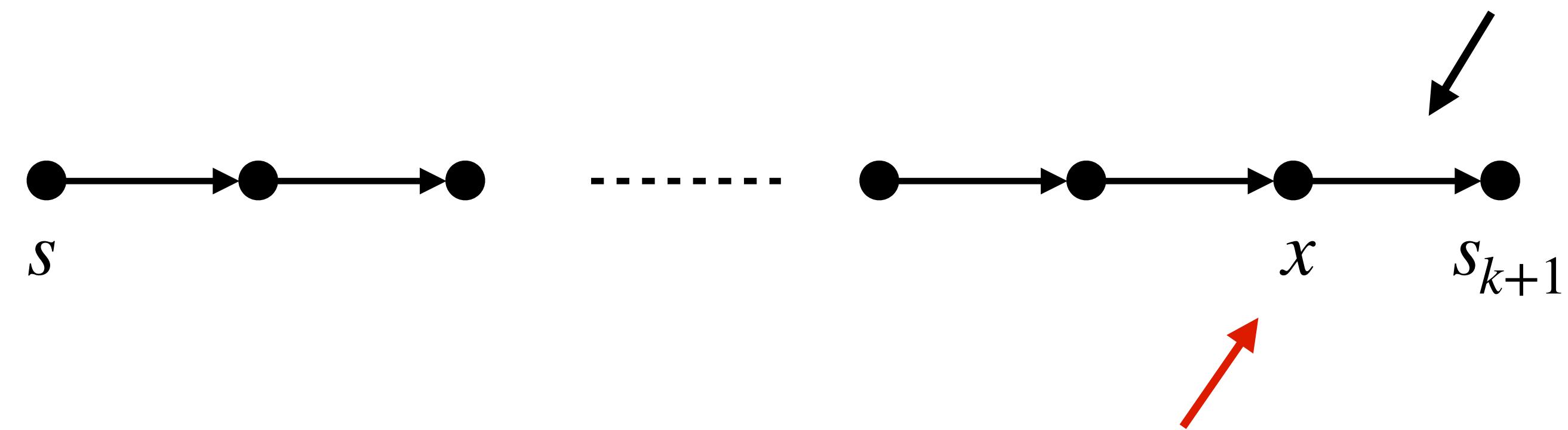
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$x$  should be  $s_i$  for some  $i \leq k$ .

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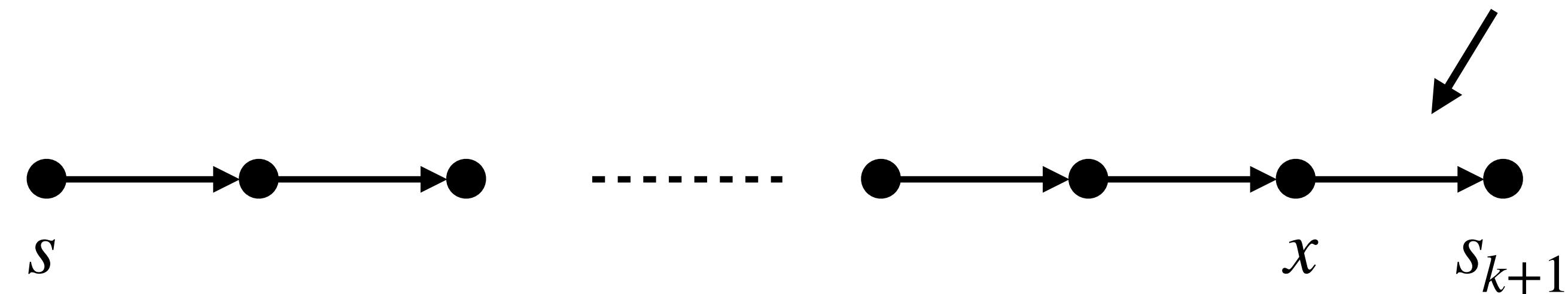
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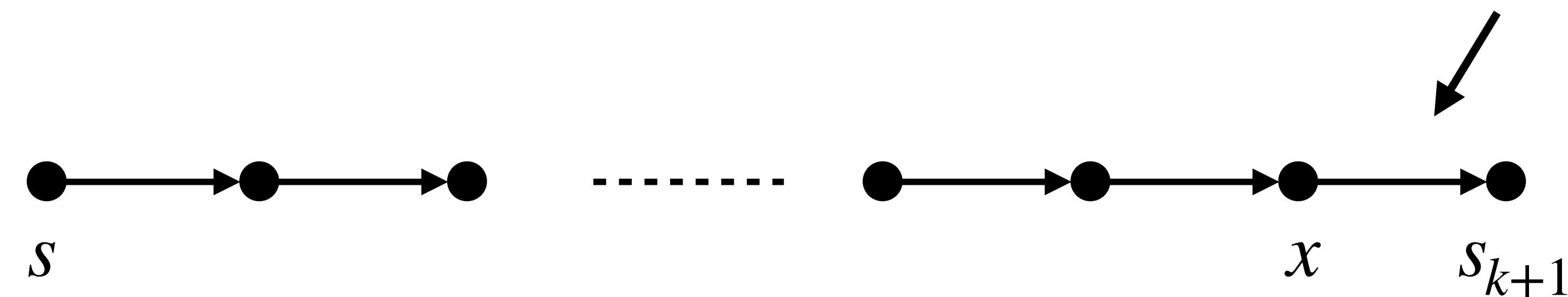
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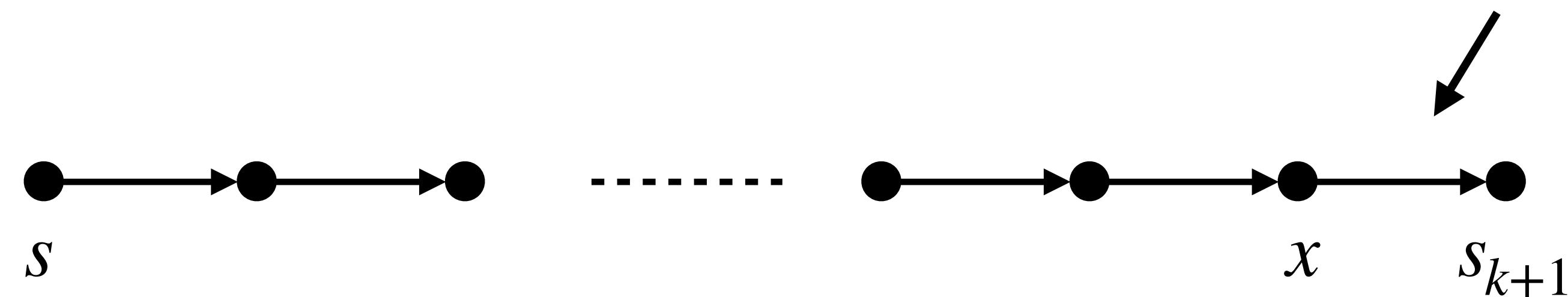
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**Step 3:** Go to **Step 2** if it can be performed.