

We can reduce computation using diagonal monotonicity:

- Whenever the value on a diagonal d grows larger than k , we can **discard** d from consideration, because we are only interested in values at most k on the row m .
- We keep track of the smallest undiscarded diagonal d . Each column is computed only up to diagonal d .

Example 4.13: $P = \text{match}$, $T = \text{remachine}$, $k = 1$

g	r	e	m	a	c	h	i	n	e
	0	0	0	0	0	0	0	0	0
m	1	1	1	0	1	1	1	1	1
a	2	2	2	1	0	1	2	2	2
t					1	1	2	3	
c						1	2	3	
h							1	2	

The position of the smallest undiscarded diagonal on the current column is kept in a variable top .

Algorithm 4.14: Ukkonen's cut-off algorithm

Input: text $T[1..n]$, pattern $P[1..m]$, and integer k

Output: end positions of all approximate occurrences of P

- (1) for $i \leftarrow 0$ to m do $g_{i0} \leftarrow i$
- (2) for $j \leftarrow 1$ to n do $g_{0j} \leftarrow 0$
- (3) $top \leftarrow \min(k + 1, m)$
- (4) for $j \leftarrow 1$ to n do
- (5) for $i \leftarrow 1$ to top do
- (6) $g_{ij} \leftarrow \min\{g_{i-1,j-1} + \delta(A[i], B[j]), g_{i-1,j} + 1, g_{i,j-1} + 1\}$
- (7) while $g_{top,j} > k$ do $top \leftarrow top - 1$
- (8) if $top = m$ then output j
- (9) else $top \leftarrow top + 1$

The time complexity is proportional to the computed area in the matrix (g_{ij}) .

- The worst case time complexity is still $\mathcal{O}(mn)$.
- The average case time complexity is $\mathcal{O}(kn)$. The proof is not trivial.

There are many other algorithms based on diagonal monotonicity. Some of them achieve $\mathcal{O}(kn)$ worst case time complexity.

Myers' Bitparallel Algorithm

Another way to speed up the computation is bitparallelism.

Instead of the matrix (g_{ij}) , we store **differences** between adjacent cells:

Vertical delta: $\Delta v_{ij} = g_{ij} - g_{i-1,j}$

Horizontal delta: $\Delta h_{ij} = g_{ij} - g_{i,j-1}$

Diagonal delta: $\Delta d_{ij} = g_{ij} - g_{i-1,j}$

Because $g_{i0} = i$ ja $g_{0j} = 0$,

$$\begin{aligned} g_{ij} &= \Delta v_{1j} + \Delta v_{2j} + \cdots + \Delta v_{ij} \\ &= i + \Delta h_{i1} + \Delta h_{i2} + \cdots + \Delta h_{ij} \end{aligned}$$

Because of diagonal monotonicity, $\Delta d_{ij} \in \{0, 1\}$ and it can be stored in one bit. By the following result, Δh_{ij} and Δv_{ij} can be stored in two bits.

Lemma 4.15: $\Delta h_{ij}, \Delta v_{ij} \in \{-1, 0, 1\}$ for every i, j that they are defined for.

The proof is left as an exercise.

Example 4.16: ‘-’ means -1, ‘=’ means 0 and ‘+’ means +1

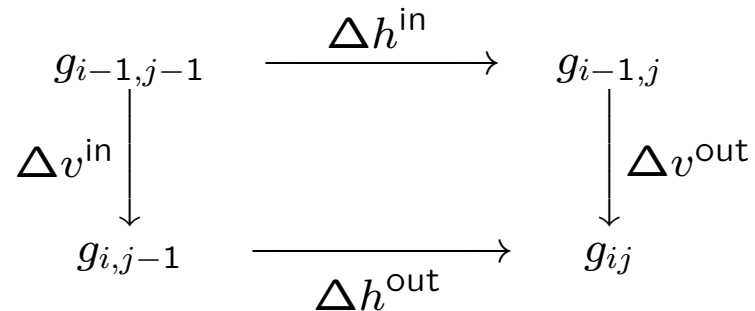
	r	e	m	a	c	h	i	n	e						
	0	=	0	=	0	=	0	=	0	=	0	=	0	=	0
m	+	+	+	+	+	=	=	+	+	+	+	+	+	+	+
	1	=	1	=	1	-	0	+	1	=	1	=	1	=	1
a	+	+	+	+	+	=	+	=	-	=	=	+	+	+	+
	2	=	2	=	2	-	1	-	0	+	1	+	2	=	2
t	+	+	+	+	+	=	+	=	+	+	=	+	+	+	+
	3	=	3	=	3	-	2	-	1	=	1	+	2	+	3
c	+	+	+	+	+	=	+	=	+	=	=	+	=	+	+
	4	=	4	=	4	-	3	-	2	-	1	+	2	+	3
h	+	+	+	+	+	=	+	=	+	=	+	=	-	=	-
	5	=	5	=	5	-	4	-	3	-	2	-	1	+	2

In the standard computation of a cell:

- Input is $g_{i-1,j}$, $g_{i-1,j-1}$, $g_{i,j-1}$ and $\delta(P[i], T[j])$.
- Output is g_{ij} .

In the corresponding bitparallel computation:

- Input is $\Delta v^{\text{in}} = \Delta v_{i,j-1}$, $\Delta h^{\text{in}} = \Delta h_{i,j-1}$ and $Eq_{ij} = 1 - \delta(P[i], T[j])$.
- Output is $\Delta v^{\text{out}} = \Delta v_{i,j}$ and $\Delta h^{\text{out}} = \Delta h_{i,j}$.



The computation rule is defined by the following result.

Lemma 4.17: If $E_q = 1$ or $\Delta v^{\text{in}} = -1$ or $\Delta h^{\text{in}} = -1$, then $\Delta d = 0$, $\Delta v^{\text{out}} = -\Delta h^{\text{in}}$ and $\Delta h^{\text{out}} = -\Delta v^{\text{in}}$. Otherwise $\Delta d = 1$, $\Delta v^{\text{out}} = 1 - \Delta h^{\text{in}}$ and $\Delta h^{\text{out}} = 1 - \Delta v^{\text{in}}$.

Proof. We can write the recurrence for g_{ij} as

$$\begin{aligned} g_{ij} &= \min\{g_{i-1,j-1} + \delta(P[i], T[j]), g_{i,j-1} + 1, g_{i-1,j} + 1\} \\ &= g_{i-1,j-1} + \min\{1 - E_q, \Delta v^{\text{in}} + 1, \Delta h^{\text{in}} + 1\}. \end{aligned}$$

Then $\Delta d = g_{ij} - g_{i-1,j-1} = \min\{1 - E_q, \Delta v^{\text{in}} + 1, \Delta h^{\text{in}} + 1\}$ which is 0 if $E_q = 1$ or $\Delta v^{\text{in}} = -1$ or $\Delta h^{\text{in}} = -1$ and 1 otherwise.

Clearly $\Delta d = \Delta v^{\text{in}} + \Delta h^{\text{out}} = \Delta h^{\text{in}} + \Delta v^{\text{out}}$. Thus $\Delta v^{\text{out}} = \Delta d - \Delta h^{\text{in}}$ and $\Delta h^{\text{out}} = \Delta d - \Delta v^{\text{in}}$. \square

To enable bitparallel operation, we need two changes:

- The Δv and Δh values are “trits” not bits. We encode each of them with two bits as follows:

$$Pv = \begin{cases} 1 & \text{if } \Delta v = +1 \\ 0 & \text{otherwise} \end{cases} \quad Mv = \begin{cases} 1 & \text{if } \Delta v = -1 \\ 0 & \text{otherwise} \end{cases}$$
$$Ph = \begin{cases} 1 & \text{if } \Delta h = +1 \\ 0 & \text{otherwise} \end{cases} \quad Mh = \begin{cases} 1 & \text{if } \Delta h = -1 \\ 0 & \text{otherwise} \end{cases}$$

Then

$$\Delta v = Pv - Mv$$
$$\Delta h = Ph - Mh$$

- We replace arithmetic operations ($+$, $-$, \min) with logical operations (\wedge , \vee , \neg).

Now the computation rules can be expressed as follows.

Lemma 4.18:

$$\begin{aligned}
 Pv^{\text{out}} &= Mh^{\text{in}} \vee \neg(Xv \vee Ph^{\text{in}}) & Mv^{\text{out}} &= Ph^{\text{in}} \wedge Xv \\
 Ph^{\text{out}} &= Mv^{\text{in}} \vee \neg(Xh \vee Pv^{\text{in}}) & Mh^{\text{out}} &= Pv^{\text{in}} \wedge Xh
 \end{aligned}$$

where $Xv = Eq \vee Mv^{\text{in}}$ and $Xh = Eq \vee Mh^{\text{in}}$.

Proof. We show the claim for Pv and Mv only. Ph and Mh are symmetrical.

By Lemma 4.17,

$$\begin{aligned}
 Pv^{\text{out}} &= (\neg\Delta d \wedge Mh^{\text{in}}) \vee (\Delta d \wedge \neg Ph^{\text{in}}) \\
 Mv^{\text{out}} &= (\neg\Delta d \wedge Ph^{\text{in}}) \vee (\Delta d \wedge 0) = \neg\Delta d \wedge Ph^{\text{in}}
 \end{aligned}$$

Because $\Delta d = \neg(Eq \vee Mv^{\text{in}} \vee Mh^{\text{in}}) = \neg(Xv \vee Mh^{\text{in}}) = \neg Xv \wedge \neg Mh^{\text{in}}$,

$$\begin{aligned}
 Pv^{\text{out}} &= ((Xv \vee Mh^{\text{in}}) \wedge Mh^{\text{in}}) \vee (\neg Xv \wedge \neg Mh^{\text{in}} \wedge \neg Ph^{\text{in}}) \\
 &= Mh^{\text{in}} \vee \neg(Xv \vee Mh^{\text{in}} \vee Ph^{\text{in}}) \\
 &= Mh^{\text{in}} \vee \neg(Xv \vee Ph^{\text{in}}) \\
 Mv^{\text{out}} &= (Xv \vee Mh^{\text{in}}) \wedge Ph^{\text{in}} = Xv \wedge Ph^{\text{in}}
 \end{aligned}$$

In the last step, we used the fact that Mh^{in} and Ph^{in} cannot be 1 simultaneously. □

According to Lemma 4.18, the bit representation of the matrix can be computed as follows.

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for  $i \leftarrow 1$  to  $m$  do
     $Pv_{i0} \leftarrow 1$ ;  $Mv_{i0} \leftarrow 0$ 
for  $j \leftarrow 1$  to  $n$  do
     $Ph_{0j} \leftarrow 0$ ;  $Mh_{0j} \leftarrow 0$ 
    for  $i \leftarrow 1$  to  $m$  do
         $Xh_{ij} \leftarrow Eq_{ij} \vee Mh_{i-1,j}$ 
         $Ph_{ij} \leftarrow Mv_{i,j-1} \vee \neg(Xh_{ij} \vee Pv_{i,j-1})$ 
         $Mh_{ij} \leftarrow Pv_{i,j-1} \wedge Xh_{ij}$ 
    for  $i \leftarrow 1$  to  $m$  do
         $Xv_{ij} \leftarrow Eq_{ij} \vee Mv_{i,j-1}$ 
         $Pv_{ij} \leftarrow Mh_{i-1,j} \vee \neg(Xv_{ij} \vee Ph_{i-1,j})$ 
         $Mv_{ij} \leftarrow Ph_{i-1,j} \wedge Xv_{ij}$ 

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This is not yet bitparallel though.

To obtain a bitparallel algorithm, the columns Pv_{*j} , Mv_{*j} , Xv_{*j} , Ph_{*j} , Mh_{*j} , Xh_{*j} and Eq_{*j} are stored in bitvectors.

Now the second inner loop can be replaced with the code

$$\begin{aligned} Xv_{*j} &\leftarrow Eq_{*j} \vee Mv_{*,j-1} \\ Pv_{*j} &\leftarrow (Mh_{*,j} \lll 1) \vee \neg(Xv_{*j} \vee (Ph_{*j} \lll 1)) \\ Mv_{*j} &\leftarrow (Ph_{*j} \lll 1) \wedge Xv_{*j} \end{aligned}$$

A similar attempt with the for first inner loop leads to a problem:

$$\begin{aligned} Xh_{*j} &\leftarrow Eq_{*j} \vee (Mh_{*j} \lll 1) \\ Ph_{*j} &\leftarrow Mv_{*,j-1} \vee \neg(Xh_{*j} \vee Pv_{*,j-1}) \\ Mh_{*j} &\leftarrow Pv_{*,j-1} \wedge Xh_{*j} \end{aligned}$$

Now the vector Mh_{*j} is used in computing Xh_{*j} before Mh_{*j} itself is computed! Changing the order does not help, because Xh_{*j} is needed to compute Mh_{*j} .

To get out of this dependency loop, we compute Xh_{*j} without Mh_{*j} using only Eq_{*j} and $Pv_{*,j-1}$ which are already available when we compute Xh_{*j} .

Lemma 4.19: $Xh_{ij} = \exists \ell \in [1, i] : Eq_{\ell j} \wedge (\forall x \in [\ell, i - 1] : Pv_{x, j-1})$.

Proof. We use induction on i .

Basis $i = 1$: The right-hand side reduces to Eq_{1j} , because $\ell = 1$. By Lemma 4.18, $Xh_{1j} = Eq_{1j} \vee Mh_{0j}$, which is Eq_{1j} because $Mh_{0j} = 0$ for all j .

Induction step: The induction assumption is that $Xh_{i-1, j}$ is as claimed. Now we have

$$\begin{aligned}
& \exists \ell \in [1, i] : Eq_{\ell j} \wedge (\forall x \in [\ell, i - 1] : Pv_{x, j-1}) \\
&= Eq_{ij} \vee \exists \ell \in [1, i - 1] : Eq_{\ell j} \wedge (\forall x \in [\ell, i - 1] : Pv_{x, j-1}) \\
&= Eq_{ij} \vee (Pv_{i-1, j-1} \wedge \exists \ell \in [1, i - 1] : Eq_{\ell j} \wedge (\forall x \in [\ell, i - 2] : Pv_{x, j-1})) \\
&= Eq_{ij} \vee (Pv_{i-1, j-1} \wedge Xh_{i-1, j}) \quad (\text{ind. assump.}) \\
&= Eq_{ij} \vee Mh_{i-1, j} \quad (\text{Lemma 4.18}) \\
&= Xh_{ij} \quad (\text{Lemma 4.18})
\end{aligned}$$

□

At first sight, we cannot use Lemma 4.19 to compute even a single bit in constant time, not to mention a whole vector Xh_{*j} . However, it can be done, but we need more bit operations:

- Let $\underline{\vee}$ denote the xor-operation: $0 \underline{\vee} 1 = 1 \underline{\vee} 0 = 1$ and $0 \underline{\vee} 0 = 1 \underline{\vee} 1 = 0$.
- A bitvector is interpreted as an integer and we use **addition** as a bit operation. The **carry** mechanism in addition plays a key role. For example $0001 + 0111 = 1000$.

In the following, for a bitvector B , we will write

$$B = B[1..m] = B[m]B[m-1] \dots B[1]$$

The reverse order of the bits reflects the interpretation as an integer.

Lemma 4.20: Denote $X = Xh_{*j}$, $E = Eq_{*j}$, $P = Pv_{*,j-1}$ ja olkoon $Y = (((E \wedge P) + P) \vee P) \vee E$. Then $X = Y$.

Proof. By Lemma 4.19, $X[i] = 1$ iff and only if

- a) $E[i] = 1$ or
- b) $\exists \ell \in [1, i] : E[\ell \dots i] = 00 \dots 01 \wedge P[\ell \dots i - 1] = 11 \dots 1$.

and $X[i] = 0$ iff and only if

- c) $E_{1\dots i} = 00 \dots 0$ or
- d) $\exists \ell \in [1, i] : E[\ell \dots i] = 00 \dots 01 \wedge P[\ell \dots i - 1] \neq 11 \dots 1$.

We prove that $Y[i] = X[i]$ in all of these cases:

- a) The definition of Y ends with “ $\vee E$ ” which ensures that $Y[i] = 1$ in this case.

b) The following calculation shows that $Y[i] = 1$ in this case:

$$\begin{array}{r}
 \\
 \\
 \\
 \\
 \\
 Y = (((E \wedge P) + P) \vee P) \vee E
 \end{array}
 \begin{array}{r}
 i \qquad \ell \\
 = 00 \dots 01 \\
 = \mathbf{b}1 \dots 11 \\
 = 00 \dots 01 \\
 = \bar{\mathbf{b}}0 \dots 0\mathbf{c} \\
 = 11 \dots 1\bar{\mathbf{c}} \\
 = 11 \dots 11
 \end{array}$$

where \mathbf{b} is the unknown bit $P[i]$, \mathbf{c} is the possible carry bit coming from the summation of bits $1 \dots, \ell - 1$, and $\bar{\mathbf{b}}$ and $\bar{\mathbf{c}}$ are their negations.

c) Because for all bitvectors B , $0 \wedge B = 0$ ja $0 + B = B$, we get
 $Y = (((0 \wedge P) + P) \vee P) \vee 0 = (P \vee P) \vee 0 = 0$.

d) Consider the calculation in case b). A key point there is that the carry bit in the summation travels from position ℓ to i and produces $\bar{\mathbf{b}}$ to position i . The difference in this case is that at least one bit $P[k]$, $\ell \leq k < i$, is zero, which stops the carry at position k . Thus $((E \wedge P) + P)[i] = \mathbf{b}$ and $Y[i] = 0$.

□

As a final detail, we compute the bottom row values g_{mj} using the equalities $g_{m0} = m$ ja $g_{mj} = g_{m,j-1} + \Delta h_{mj}$.

Algorithm 4.21: Myers' bitparallel algorithm

Input: text $T[1..n]$, pattern $P[1..m]$, and integer k

Output: end positions of all approximate occurrences of P

- (1) for $c \in \Sigma$ do $B[c] \leftarrow 0^m$
- (2) for $i \leftarrow 1$ to m do $B[P[i]][i] = 1$
- (3) $Pv \leftarrow 1^m$; $Mv \leftarrow 0$; $g \leftarrow m$
- (4) for $j \leftarrow 1$ to n do
- (5) $Eq \leftarrow B[T[j]]$
- (6) $Xh \leftarrow (((Eq \wedge Pv) + Pv) \underline{\vee} Pv) \vee Eq$
- (7) $Ph \leftarrow Mv \vee \neg(Xh \vee Pv)$
- (8) $Mh \leftarrow Pv \wedge Xh$
- (9) $Xv \leftarrow Eq \vee Mv$
- (10) $Pv \leftarrow (Mh \ll 1) \vee \neg(Xv \vee (Ph \ll 1))$
- (11) $Mv \leftarrow (Ph \ll 1) \wedge Xv$
- (12) $g \leftarrow g + Ph[m] - Mh[m]$
- (13) if $g \leq k$ then output j