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: $P =$ `match`, $T =$ `remachine`, $k = 1$

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

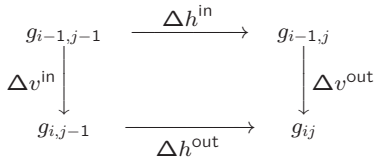
	r	e	m	a	c	h	i	n	e
	0 = 0 = 0 = 0 = 0 = 0 = 0 = 0 = 0								
m	+ + + + = = + + + + + + + + +								
	1 = 1 = 1 − 0 + 1 = 1 = 1 = 1 = 1								
a	+ + + + + = + = − = = + + + + + + +								
	2 = 2 = 2 − 1 − 0 + 1 + 2 = 2 = 2 = 2								
t	+ + + + + = + = + + = + = + + + + +								
	3 = 3 = 3 − 2 − 1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

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 $Eq = 1$ 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

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

Then $\Delta d = g_{ij} - g_{i-1,j-1} = \min\{1 - Eq, \Delta v^{\text{in}} + 1, \Delta h^{\text{in}} + 1\}$
which is 0 if $Eq = 1$ 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 $\Delta h^{\text{out}} = \Delta d - \Delta v^{\text{in}}$. □

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:

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

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

  Then

$$\begin{aligned} \Delta v &= Pv - Mv \\ \Delta h &= Ph - Mh \end{aligned}$$

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

Now the computation rules can be expressed as follows.

Lemma 4.18:
$$Pv^{\text{out}} = Mh^{\text{in}} \vee \neg(Xv \vee Ph^{\text{in}}) \qquad Mv^{\text{out}} = Ph^{\text{in}} \wedge Xv$$
$$Ph^{\text{out}} = Mv^{\text{in}} \vee \neg(Xh \vee Pv^{\text{in}}) \qquad Mh^{\text{out}} = Pv^{\text{in}} \wedge Xh$$
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,

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

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.

```
for i ← 1 to m do
    Pv_i0 ← 1; Mv_i0 ← 0
for j ← 1 to n do
    Ph_0j ← 0; Mh_0j ← 0
    for i ← 1 to m do
        Xh_ij ← Eq_ij ∨ Mh_{i-1,j}
        Ph_ij ← Mv_{i,j-1} ∨ ¬(Xh_ij ∨ Pv_{i,j-1})
        Mh_ij ← Pv_{i,j-1} ∧ Xh_ij
    for i ← 1 to m do
        Xv_ij ← Eq_ij ∨ Mv_{i,j-1}
        Pv_ij ← Mh_{i-1,j} ∨ ¬(Xv_ij ∨ Ph_{i-1,j})
        Mv_ij ← Ph_{i-1,j} ∧ Xv_ij
```

This is not yet bit*parallel* 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

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

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

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

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

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

□

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 $\veebar$ denote the xor-operation: $0 \veebar 1 = 1 \veebar 0 = 1$ and $0 \veebar 0 = 1 \veebar 1 = 0$.
- 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

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

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 $Y = (((E \wedge P) + P) \veebar P) \vee E$. Then $X = Y$.

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

a) $E[i] = 1$ or

b) $\exists \ell \in [1,i] : E[\ell \ldots i] = 00\cdots01 \wedge P[\ell \ldots i-1] = 11\cdots1$.

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

c) $E_{1\ldots i} = 00\cdots0$ or

d) $\exists \ell \in [1,i] : E[\ell \ldots i] = 00\cdots01 \wedge P[\ell \ldots i-1] \neq 11\cdots1$.

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:

$$\begin{aligned} & & i & & \ell \\ E[\ell \ldots i] &= 00 & \ldots & & 01 \\ P[\ell \ldots i] &= \mathtt{b}1 & \ldots & & 11 \\ (E \wedge P)[\ell \ldots i] &= 00 & \ldots & & 01 \\ ((E \wedge P) + P)[\ell \ldots i] &= \bar{\mathtt{b}}0 & \ldots & & 0\mathtt{c} \\ (((E \wedge P) + P) \veebar P)[\ell \ldots i] &= 11 & \ldots & & 1\bar{\mathtt{c}} \\ Y = ((((E \wedge P) + P) \veebar P) \vee E)[\ell \ldots i] &= 11 & \ldots & & 11 \end{aligned}$$

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

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

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 $\bar{\mathtt{b}}$ 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 $((E \wedge P) + P)[i] = \mathtt{b}$ 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 ∈ Σ do B[c] ← 0^m
(2)  for i ← 1 to m do B[P[i]][i] = 1
(3)  Pv ← 1^m; Mv ← 0; g ← m
(4)  for j ← 1 to n do
(5)      Eq ← B[T[j]]
(6)      Xh ← (((Eq ∧ Pv) + Pv) ⊻ Pv) ∨ Eq
(7)      Ph ← Mv ∨ ¬(Xh ∨ Pv)
(8)      Mh ← Pv ∧ Xh
(9)      Xv ← Eq ∨ Mv
(10)     Pv ← (Mh << 1) ∨ ¬(Xv ∨ (Ph << 1))
(11)     Mv ← (Ph << 1) ∧ Xv
(12)     g ← g + Ph[m] − Mh[m]
(13)     if g ≤ k then output j
```