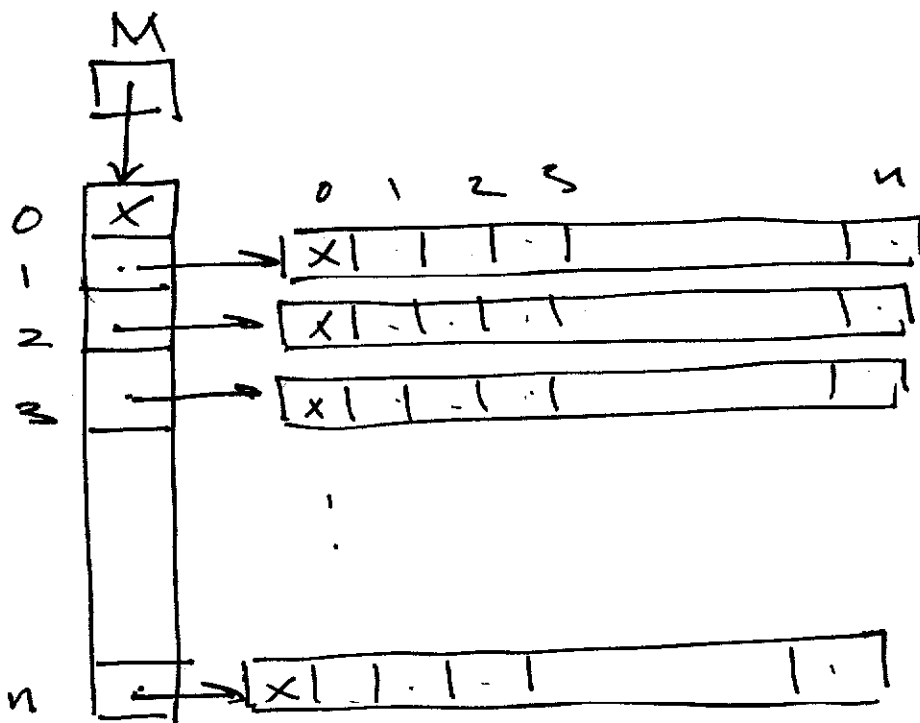


o Paz ext. 1 last day

Paz:

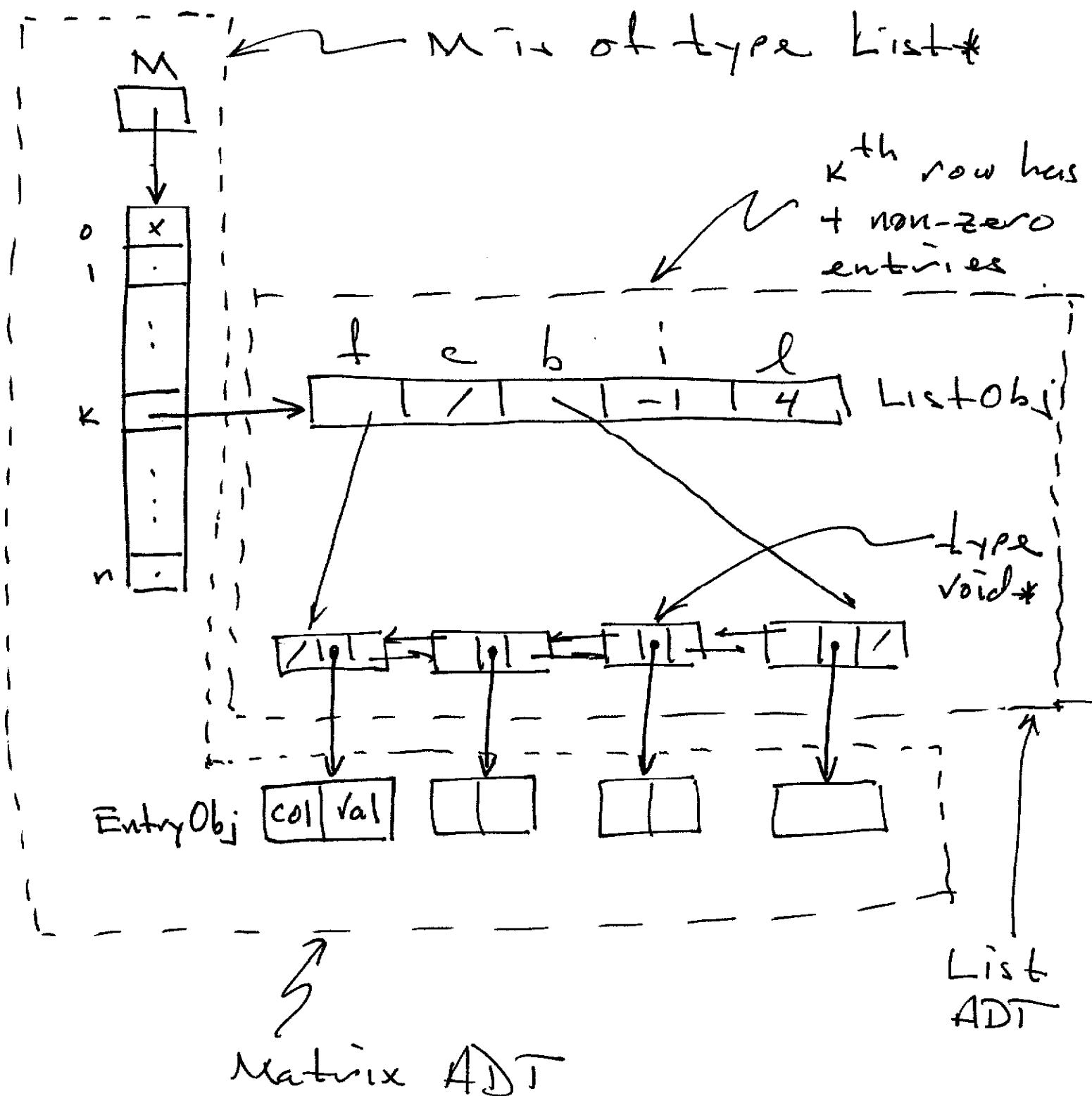
traditional matrix representation!



n would have type `double**`

we don't use this matrix representation.

in Part 4 we use following sparse matrix representation



Algorithm Runtime & Efficiency

To measure runtime of an (iterative) algorithm, first establish some basic operation(s), and count # of times those ops. are performed.

Ex.

```

OP1
for i = 1 to 10
    OP2
    for j = 1 to 10
        OP3
        for k = 1 to 10
            OP4
  
```

OP1	1	} # of executions
OP2	10	
OP3	100	
OP4	1000	

← Count only this one

Assume $OP_1 \dots OP_4$ are approx.
equal cost.

- (1) what if they're not equal?
- (2) why not count all?

(1) re-write algorithm in terms
of (1) basic OP.

new algorithm:

```

OP.
for i = 1 to 10
  OP.
  for j = 1 to 10
    OP
    for k = 1 to 10
      OP.
  
```

more common to have

```

OP
for i = 1 to n
    OP
    for j = 1 to n
        OP
        for k = 1 to n
            OP

```

(# of ops as a fun. of input n)

$$= T(n) = n^3 + n^2 + n + 1$$

we propose to replace this by n^3 ,
i.e. only count innermost loop.

①

```

for i = 1 to n
    for j = 1 to n
        for k = 1 to n
            OP

```

Question we're interested in
is how runtime grows with input
size.

compare using limits:

$$\begin{aligned}\lim_{n \rightarrow \infty} \left(\frac{T(n)}{n^3} \right) &= \lim_{n \rightarrow \infty} \frac{(n^3 + n^2 + n + 1)}{n^3} \\ &= \lim_{n \rightarrow \infty} \left(1 + \frac{1}{n} + \frac{1}{n^2} + \frac{1}{n^3} \right) \\ &\quad \begin{array}{ccc} \downarrow & \downarrow & \downarrow \\ 0 & 0 & 0 \end{array} \\ &= \boxed{1}\end{aligned}$$

so we can ignore lower order
terms

consider another algorithm

②

```

for i = 1 to n
  for j = 1 to n
    for k = 1 to n
      OP
      OP
  
```

its runtime is : $2n^3$

Assume ① and ② accomplish the same thing, so we can compare them.

compare : $\underbrace{n^3}_{\text{twice as fast}}$ to $\underbrace{2n^3}$

we consider these equivalent!!

why?

18

we can equalize them by
running (2) on a machine that
is twice as fast.

It not always possible to
do this. consider

(3)

for $i = 1$ to n
 for $j = 1$ to n
 for $k = 1$ to 10
 op

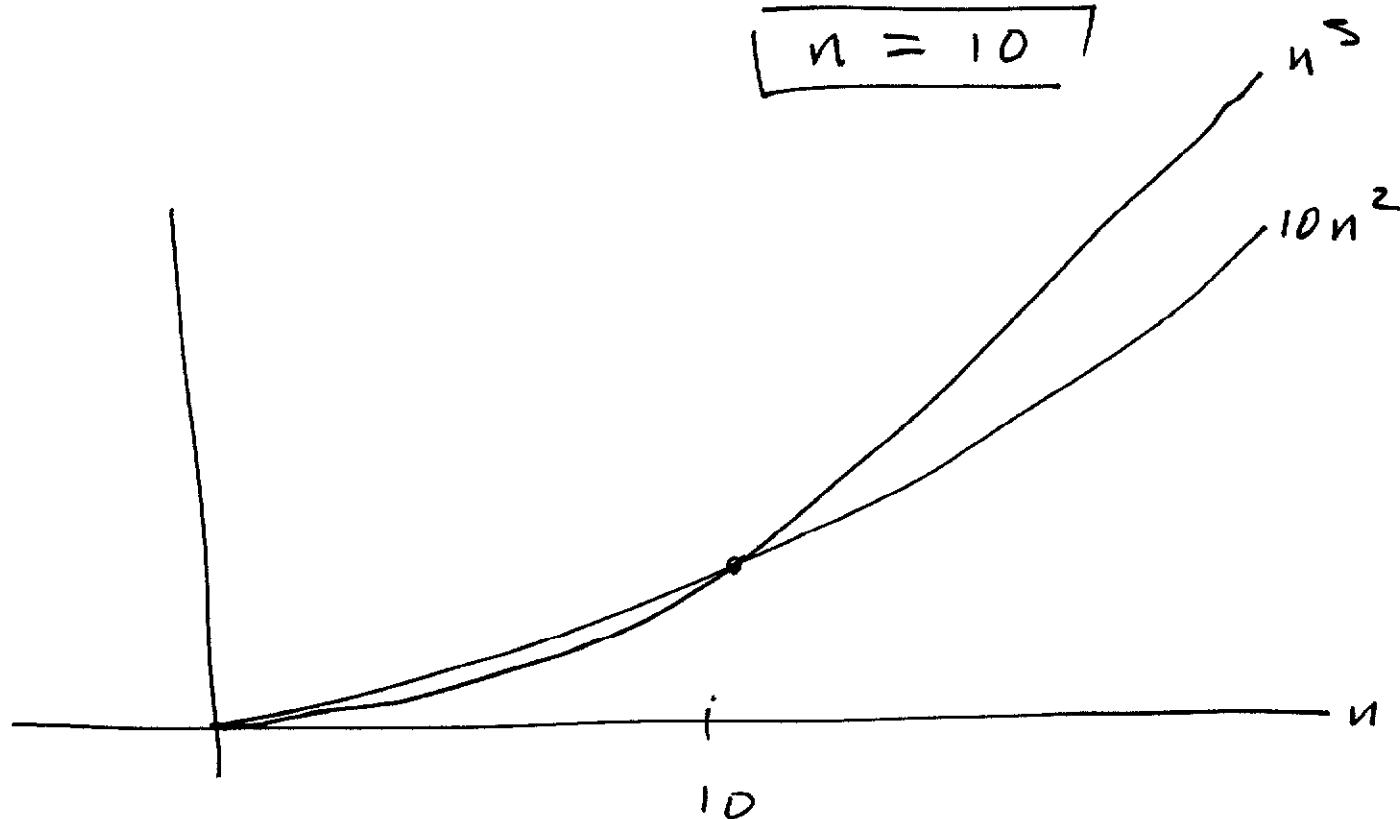
$$\text{runtime} = 10n^2$$

Compare (1) to (3)

$$n^3 \text{ to } 10n^2$$

Crossover Point: $n^3 = 10n^2$

$$n = 10$$



in the limit $n \rightarrow \infty$, $10n^2$ is smaller;

$$\frac{10n^2}{n^3} = \frac{10}{n} \rightarrow 0$$

so $\textcircled{3}$ is a superior algorithm.

Informal method :

(1) choose a basic op. inside the body of innermost loop.

(2) count # of execution of op as a function of input size n : $f(n)$

(3) simplify then $f(n)$ by

- drop lower order terms

- replace coeff. of highest order term by 1.

Result : asymptotic runtime

Asymptotic growth of functions

Let $f(n), g(n)$ be functions.

Defn

write $f(n) = O(g(n))$ to mean

there exists $R > 0$ and $n_0 > 0$
such that $\frac{f(n)}{g(n)} \leq R$ for all $n \geq n_0$

We say

- $g(n)$ is an asymptotic upper bound for $f(n)$
- $f(n)$ is asymptotically bounded above
by $g(n)$

means: $f(n)$ grows no faster than $g(n)$.