



# Study Group - Lect #4

## Mixtures of Permutations [Mao, Wu, 2020].

$$\pi \in \mathbb{S}_n$$

- 1) Perm. matrix
- 2) Clique (directed)  $\leadsto$  Tournaments.
- 3) ...

Statistical models on permutations.

### Motivation

Netflix  $N$  movies  $\rightarrow$   $[m_1 > m_2 > m_3 > m_5 > m_4]$

Mallows model  $(\pi_0, \phi)$   
 $\pi_0 \in \mathbb{S}_n$   
 distance central ranking

$$\Pr[\sigma | \pi_0, \phi] \propto \phi^{KT(\pi_0, \sigma)}$$

$$0 < \phi < 1$$

$\pi_0: 12345$

$\pi_1: 21345$

$\pi_2: 54321$

$KT(\pi, \sigma) = \frac{\# \text{steps}}{\text{bubblesort}}$   
 should make starting from  $\pi$  ending at  $\sigma$ .

$$KT(\pi, \sigma) = \sum_{i < j} \mathbb{1}\{(\pi(i) - \pi(j))(\sigma(i) - \sigma(j)) < 0\}$$

Physics  
 $\phi = e^{-\beta}$ ,  $\beta > 0$   
 inverse temperature.

if  $n$  is closer to  $n_0$   
than  $\sigma$  then  $\Pr[n|n_0, \phi] > \Pr[\sigma|n_0, \phi]$ .

$$\Pr[n|n_0, \phi] \propto \phi^{KT(n_0, n)}$$

## Mixtures of Mallows

$$\mu = \sum w_i \mu(n_0^{(i)}, \phi)$$

Retrieve  $\{n_0^{(i)}\}$ ,  $\boxed{(w_i)}$

## Identifiability

Zagier

$$A(\phi) \in \mathbb{R}^{n! \times n!}$$

$$= \left[ \phi^{KT(n, \sigma)} \right]_{n, \sigma}$$

symmetric ✓

PMF of  $\mu(n, \phi)$

$$\det(A(\phi)) \neq 0 \quad \forall \phi \in (0, 1)$$

$$\mu = \sum w_i \mu(\pi_i, \phi)$$

Draw

$$\sigma_1, \sigma_2, \dots, \sigma_N$$

→

Can you learn  $(\pi_i)$ ?

Try something simpler:

$$\boxed{\sigma_n}$$

Oracle

output

n

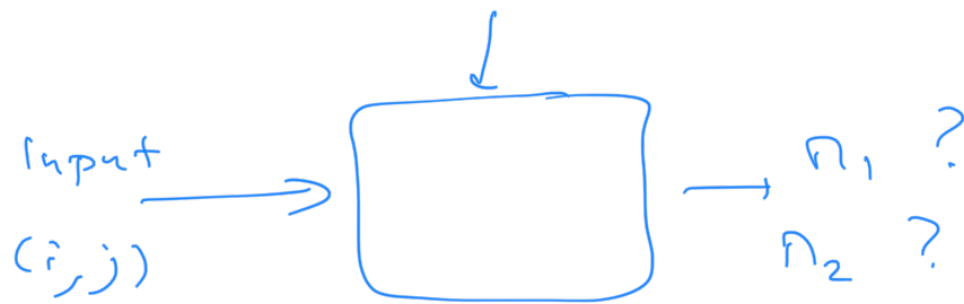
...

n



Algo: Quicksort.

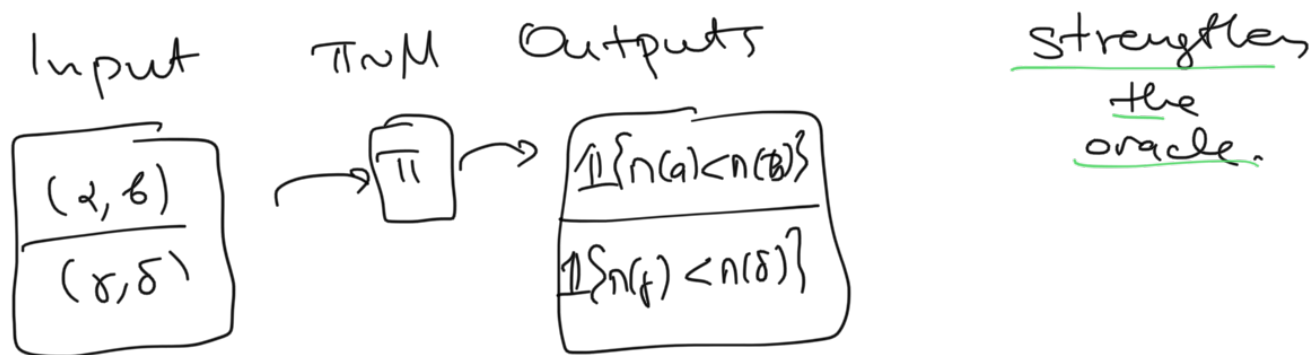
Example 2:  $\frac{1}{2} (\delta n_1 + \delta n_2)$



Pairwise Comp. Oracle is not sufficient.  
 for identifying  $M$ .

$$\underline{\pi_1} = \underline{\pi_2^{-1}} \Rightarrow \text{Be}(\frac{1}{2})$$

Example 3 : You can compare two pairs simultaneously

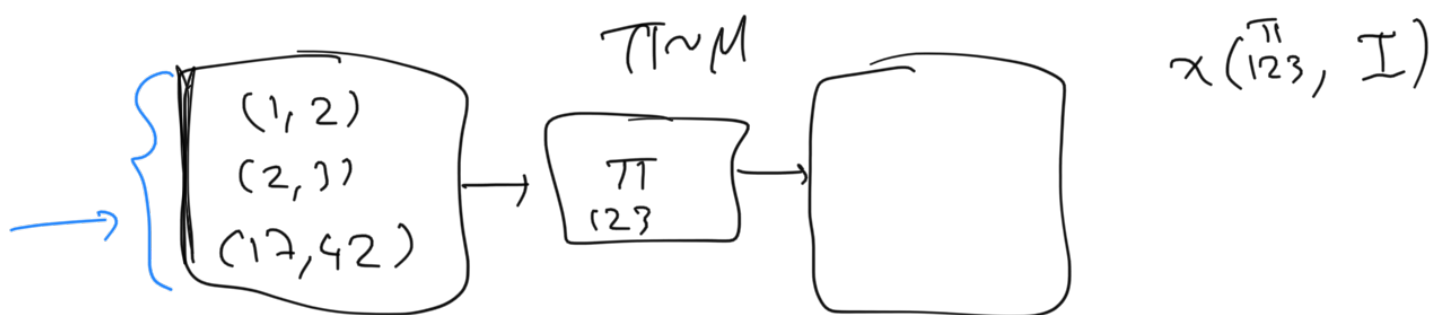
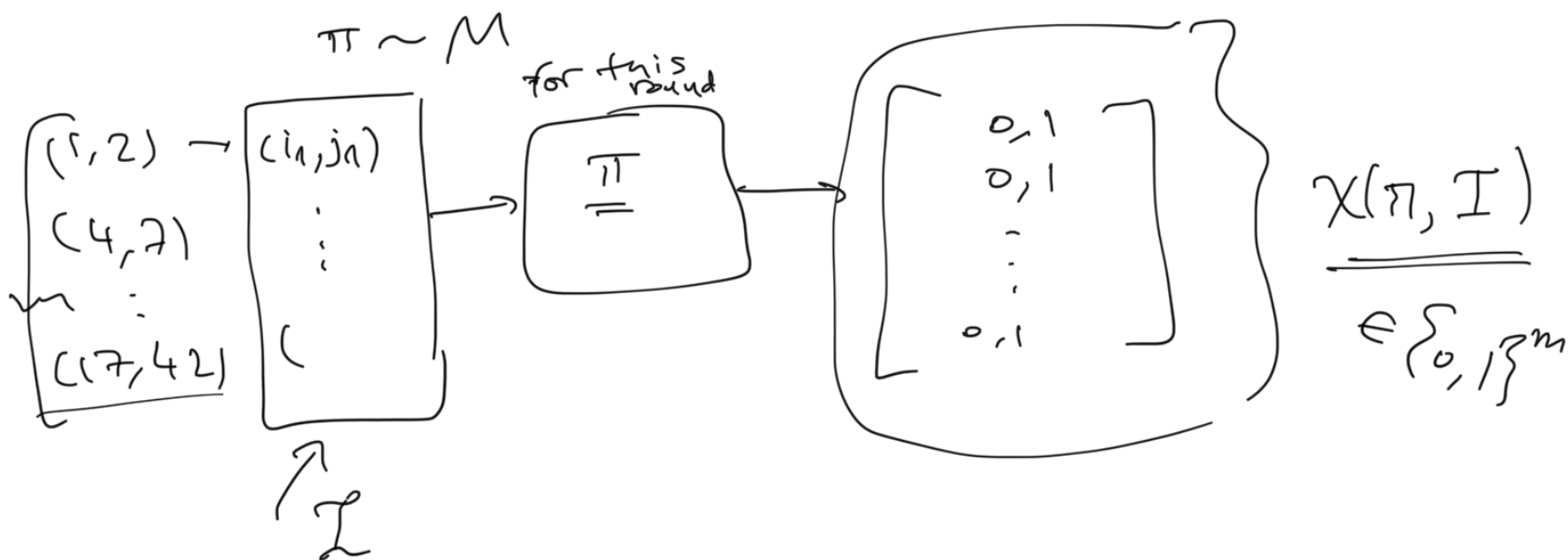


$\frac{1}{2}(\delta_{n_1} + \delta_{n_2})$  where  $\frac{1 <_{n_1} 2}{1 >_{n_2} 2}$



Oracle

Group of pairwise comparisons, size  $m$



This is the (strong) oracle.

Noiseless case ( $\phi=0$ )

$$M = \sum_{i=1}^k w_i \delta_{\pi_i} \quad \pi_i \in \Sigma_n$$

Q: Is there a poly-time algo  
that learns the  $k$ -mixt  $M$   
from a group of poly-sized  
 $m$  (pairwise) comparisons for any large  $n$

where  $m$  depends only on  $k$

Thm:  $m = ???$

$$m = f(k) = \lfloor \log_2 k \rfloor + 1$$

Let  $m_k^*$   $= \lfloor \log_2 k \rfloor + 1$

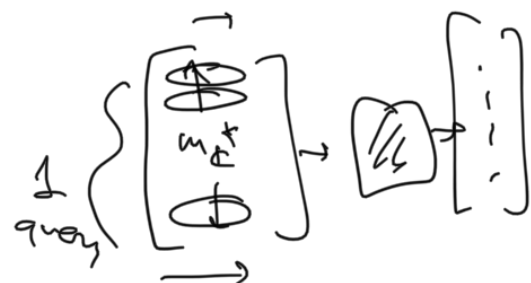
For any  $M = \sum_{i=1}^k w_i \delta_{\pi_i}$ ,  $\exists$   $\text{poly}(n, k)$ -time

algo that recovers  $M$  from gps

of  $m_k^*$  pairwise

comps with

$O\left(\frac{k}{2}(n-2)(n+1)\right)$  adaptive queries.



no comment

(to the weak oracle)

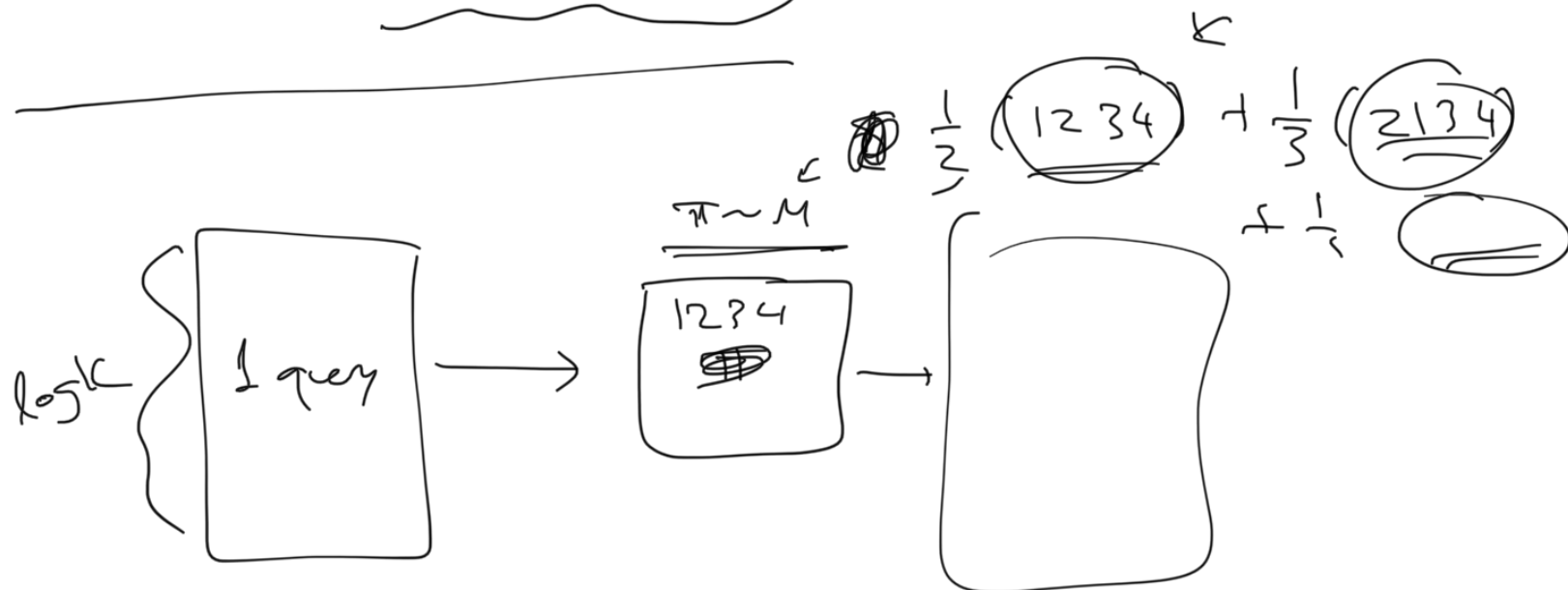
Key lemma.

Give me  $k$  perms  $\Sigma = \{\pi_1, \dots, \pi_k\}$  in  $S_n$

$\exists \pi^* \in \Sigma$  and  $I$  of  $l$  pairs  
 $\uparrow$   
 $O(\log k)$

~~s.t.~~ s.t.

$$\chi(\pi^*, I) \neq \chi(\pi, I) \quad \forall \pi \in \Sigma \setminus \{\pi^*\}$$



Decomposing Mallows

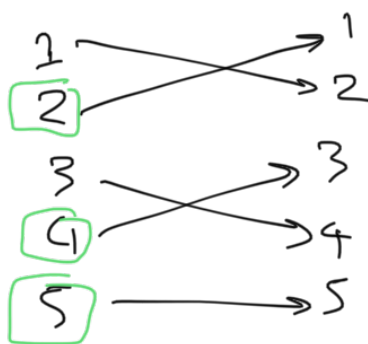
$J \subset [n]$

$\pi|_J$

$\pi$

$(J)$

$\pi||_J$



$\pi|_J$  : injection (1-1)

$$\pi|_J(2) = 1$$

$$\pi|_J(4) = 3$$

$$\pi|_J(5) = 5 \quad \checkmark$$

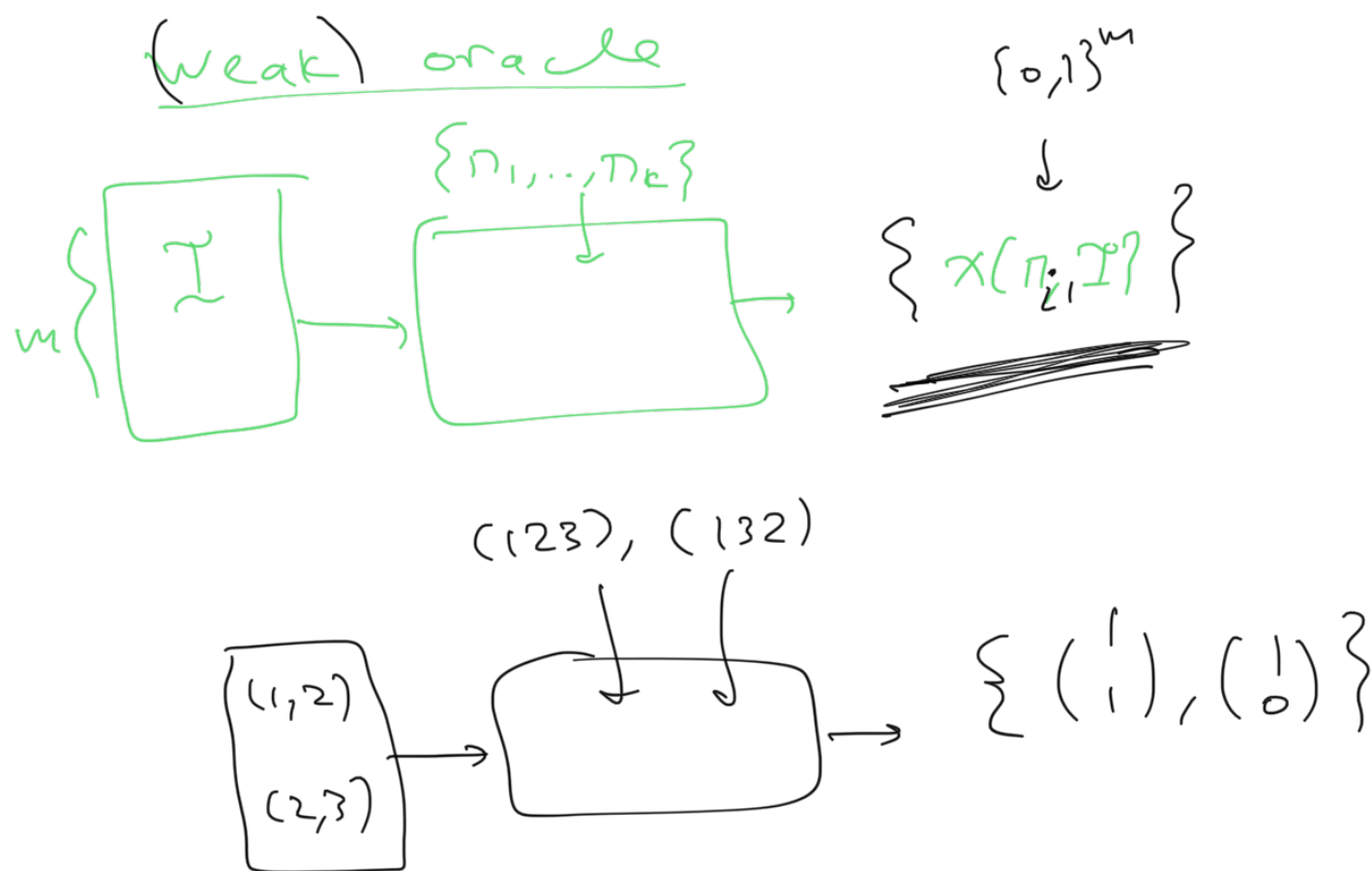
$=$

$$\left\{ \begin{array}{l} \pi||_J \in S_{|J|} \\ \pi||_J(2) = 1 \\ \pi||_J(4) = 2 \\ \pi||_J(5) = 3 \end{array} \right\} \in S_3$$

$\uparrow$   
 $=$   
 $\pi$

Subroutine ( $\mathcal{J}$ )  $\mathcal{J} \subseteq [n]$   
 Given  $\sigma_1, \dots, \sigma_N \sim M = \sum w_i \underline{M(\pi_i)}$   
 the subroutine ( $\mathcal{J}$ ) learns  
 the set  
 $\{ \underline{\pi_1 \parallel_{\mathcal{J}}}, \dots, \underline{\pi_k \parallel_{\mathcal{J}}} \} \rightarrow$  whp  
output of  
the  
weak  
oracle

Assume that we have this algo



$(i,j) \in \mathcal{J} \times \mathcal{J}$  Smart idea

$$\mathbb{1} \{ \pi \parallel_{\mathcal{J}}(i) < \pi \parallel_{\mathcal{J}}(j) \}$$

$$\mathbb{1} \{ \pi(i) < \pi(j) \}$$

(1) Subroutine

(2) Moments (High-noise regime  $c \rightarrow 1$ ) ( $\leftarrow$ )



$$(1) \quad \{\sigma_i\}_{i=1}^N \sim M = \sum w_i M(\pi_i)$$

goal:  $\pi_i \|_J$   $(J)$

$\rho$  injection  $\underline{\underline{\sigma|_J}}$

$$f(\rho) = \Pr[\sigma|_J = \rho]$$

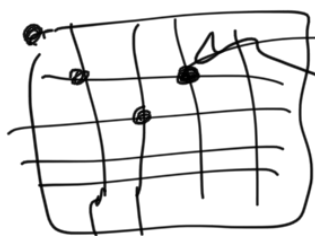
$= \sum_{\sigma \sim M}$

Design empirical

$$f_{\text{emp}}(\rho) = \frac{1}{N} \sum \mathbb{1}\{\sigma_m|_J = \rho\}$$

Idea: Collect poly # of Mallas mixtures & you do grid search

$$\mathcal{M} = \left\{ \sum_{i=1}^K \begin{pmatrix} r_i \\ L \end{pmatrix} M(\underline{\pi}_{\rho_i}) \right\}$$



$(N)$  emp.

Compare

$$TV(\text{emp in the grid}, \text{emp of the true}) \leq \gamma$$

Return

$$\{\pi_{\rho_i}\|_J\}$$

feed this thing to the TM for

noiselen.

(2) Moment

High-noise

$\phi \rightarrow 1$

$$\underline{\varepsilon = 1 - \phi}$$

$$M = \sum_{i=1}^k \frac{1}{k} M(\pi_i)$$

$$M' = \sum_{i=1}^k \frac{1}{k} M(\pi'_i)$$

$$TV(M, M') = ?$$

$$f(\sigma) = \frac{1}{Z} (1 - \varepsilon)^{k\tau(\pi_i, \sigma)}$$

$$= \frac{1}{Z} \sum_{l=0}^{k\tau(\sigma, \pi_i)} \binom{k\tau(\sigma, \pi_i)}{l} (-\varepsilon)^l$$

Thm

$$TV(M, M') = O(\varepsilon^{\frac{1}{k+1}})$$

$$\sum_{i=1}^k k\tau(\sigma, \pi_i)^l = \sum_{i=1}^k k\tau(\sigma, \pi'_i)^l$$

$\forall l \in [n], \sigma \in \mathcal{S}_n$

moment-like thing

Informal

$$\sum x^r$$

$$k\tau(\pi, \sigma) = \sum_{\substack{(i,j) \\ \sigma(i) > \sigma(j)}} \mathbb{1}\{\pi(i) < \pi(j)\}$$

$$|KT(\sigma, \pi)|$$

$$KT(\pi, \sigma) = \sum_{(i_1, j_1), \dots, (i_\ell, j_\ell)} \prod_{k=1}^{\ell} \pi(X_{i_k j_k}^{\pi})$$

$$\pi \quad \sigma$$

$$\mathbb{1}\{\pi(i) < \pi(j)\}$$

$$X_{i,j}^{\pi}$$

$$KT(\pi, \sigma) = \sum_{(i,j)} X_{i,j}^{\pi}$$

$$\sigma(i) > \sigma(j)$$

$$KT(\pi, \sigma)^2 = \left( \sum_{(i,j)} X_{i,j}^{\pi} \right)^2$$

$$= \left( \sum_{\substack{(i,j) \\ \sigma(i) > \sigma(j)}} X_{i,j}^{\pi} \right) \left( \sum_{\substack{(i,j) \\ \sigma(i) > \sigma(j)}} X_{i,j}^{\pi} \right)$$

$$= \sum_{\substack{(i,j) (a,b) \\ \sigma(i) > \sigma(j) \\ \sigma(a) > \sigma(b)}} X_{i,j}^{\pi} X_{a,b}^{\pi}$$

$$\sum x^r$$

$$\pi \rightarrow \text{exp. large natural}$$

$$C_{i,j}$$

Moment

Problems  
⋮

Hamburger  
⋮

Hausarbeit  
⋮

Stilligen.  
⋮

$\mu \in$  moments

$\left[ \begin{array}{c} \text{power} \\ \text{comp} \end{array} \right]$   
↓  
power  
of ↗

$\left[ \Pi \right]$