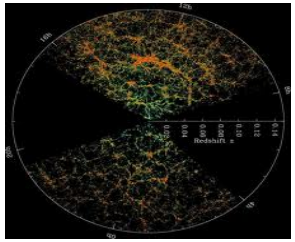
Distributed data mining in accessing the data from VO's

Downloaded Table

- All the columns of the downloaded table may not be relevant
- Some of the columns may be redundant .

http://en.wikipedia.org/wiki/Fundamental_plane_(elliptical_galaxies)

Web services



Sky survey I

Sky survey II

Sky survey III

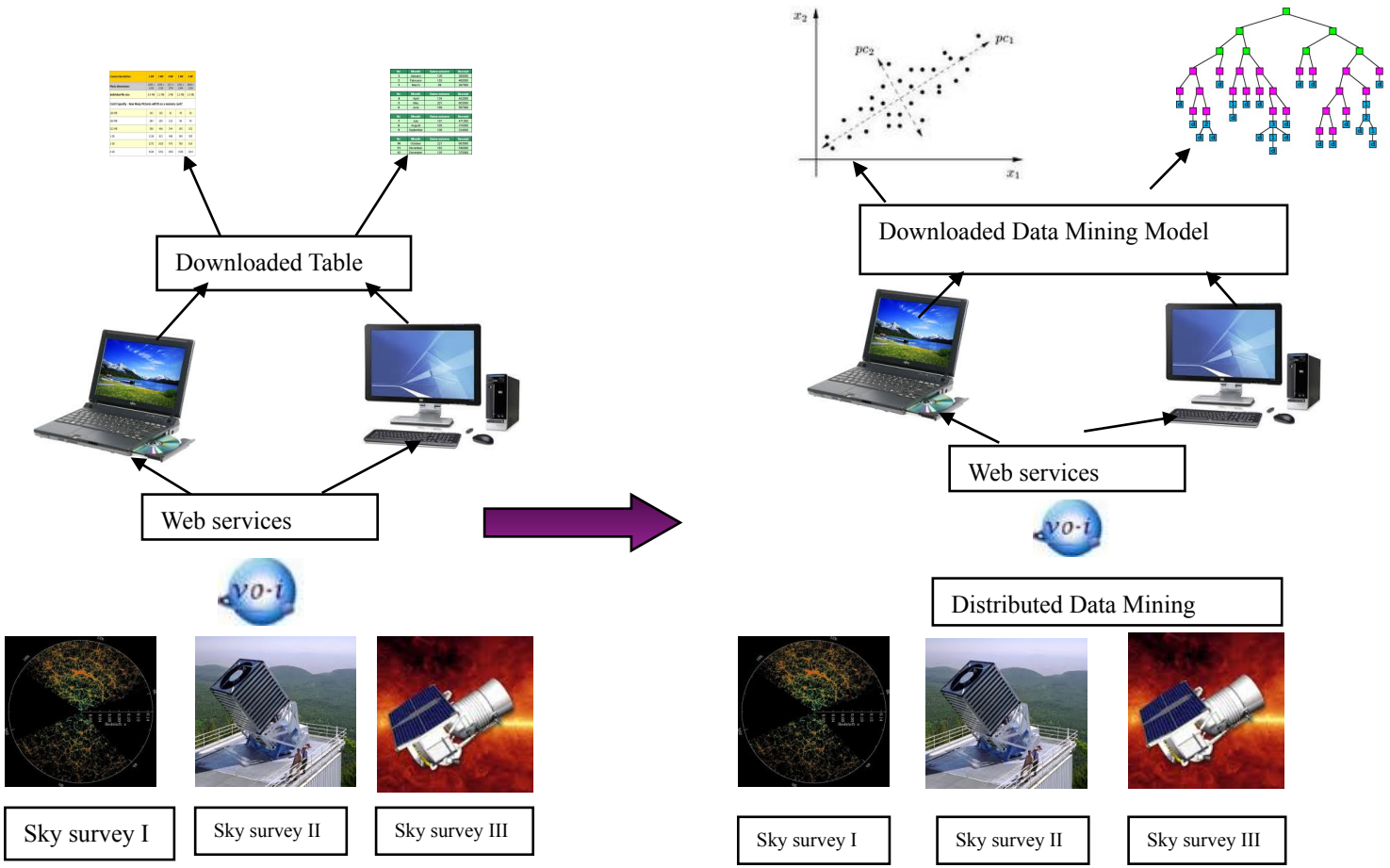
For Eg:

The fundamental plane is a relationship between the effective radius, average surface brightness and central velocity dispersion of normal elliptical galaxies.

What can be done ??

- **We can embed the service of filtering the data by applying the data mining algorithms and provide the data mining model instead of raw tables.**

- **Which needs DDM (Distributed Data Mining) to be carried out without having to down-load large tables to the user.**

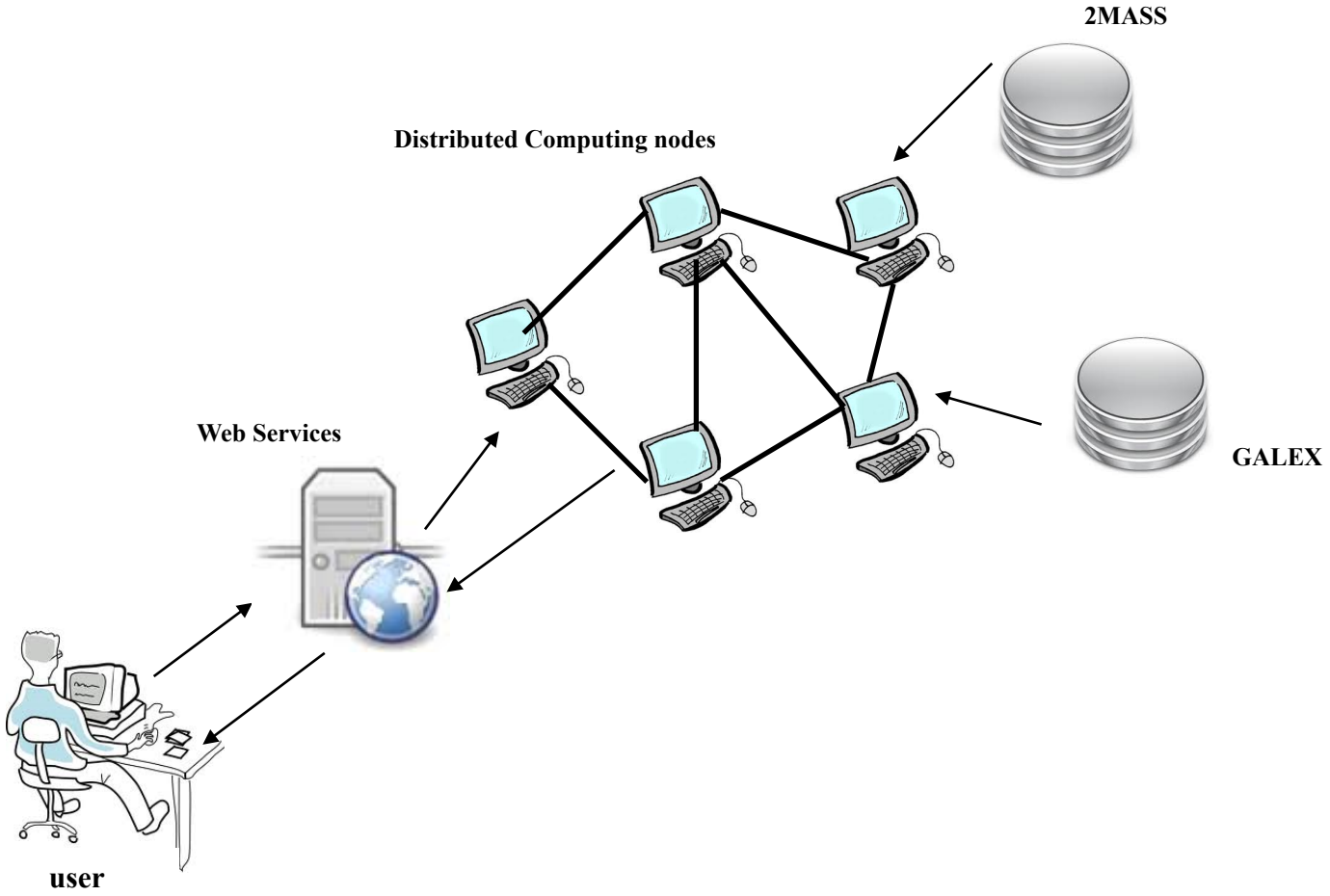


a. Users are getting only raw data

b. Users can get data mining model rather than raw data

rather than raw data

Fig 1 . Distributed Data Mining Data Flow that can be embedded in VO-I


2MASS
Distributed Computing nodes
Web Services
GALEX
user

DDM (Distributed Data Mining)

- DDM strive to analyze the data in a distributed manner without down-loading all the data to a single site.
- DDM is possible in horizontal or vertical partitions .
- In case of horizontal the data is divided among rows, but the number of columns are same at all sites.
- Where as in vertical partition the data is divided among columns ,but the number of rows are same at all sites.
- We considered vertical partition for our implementation .

As an initial step….

- **Reducing the dimension of large high-dimensional data sets will make the analysis efficient .**
- **Reduction of dimensionality using principal component analysis.**
- **PCA can be computed from eigen vectors of covariance matrix.**
- **In our implementation covariance matrix is calculated in a distributed manner.**

Distributed Principal Component Analysis

Problem:

Data are distributed (vertically partitioned) amongst t nodes .

$$[X]_{n \times m} = (X_0 X_1 X_2 X_3 X_4 \ldots\ldots\ldots X_{t-1})$$

where Xj resides at node Sj ,
a $n \times mj$ matrix , $\sum j=1 \text{ to } t \; mj = m$

Aim:

Compute PCA of X without moving X ($X_0 X_1 X_2 X_3 X_4 \ldots\ldots\ldots X_{t-1}$) data matrix to a central location such that to avoid the communication and computation bottleneck.

Demonstration with 3 nodes

For example the status of the data is as follows

- node 0 ------x y columns
- node 1 ------z w columns
- node 2 ------l column

•The data need not be centralized like….

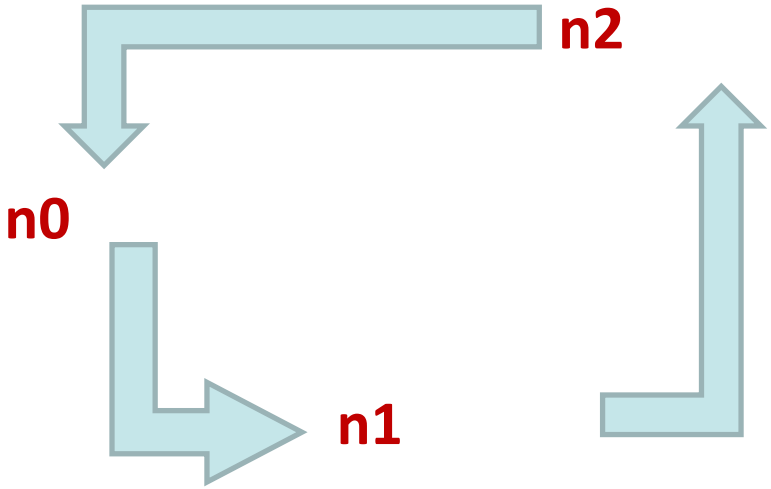
$$\begin{matrix} x_1 & y_1 & z_1 & w_1 & l_1 \\ x_2 & y_2 & z_2 & w_2 & l_2 \\ x_3 & y_3 & z_3 & w_3 & l_3 \\ . & . & . & . & . \\ . & . & . & . & . \\ . & . & . & . & . \\ x_m & y_m & z_m & w_m & l_m \end{matrix}$$

To calculate covariance matrix

$$\begin{matrix} xx & xy & xz & xw & xl \\ yx & yy & yz & yw & yl \\ zx & zy & zz & zw & zl \\ wx & wy & wz & ww & wl \\ lx & ly & lz & lw & ll \end{matrix}$$

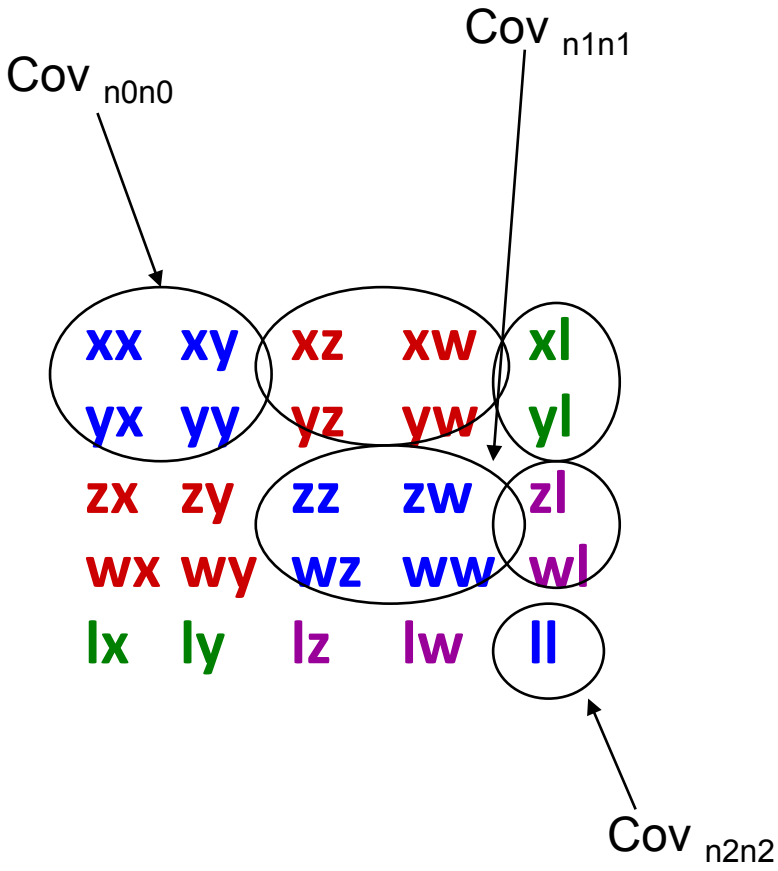
The communication b/w 3 nodes

1.sends data to n0
2.Calculates Cov $_{n2n2}$
3.Calculates Cov $_{n1n2}$

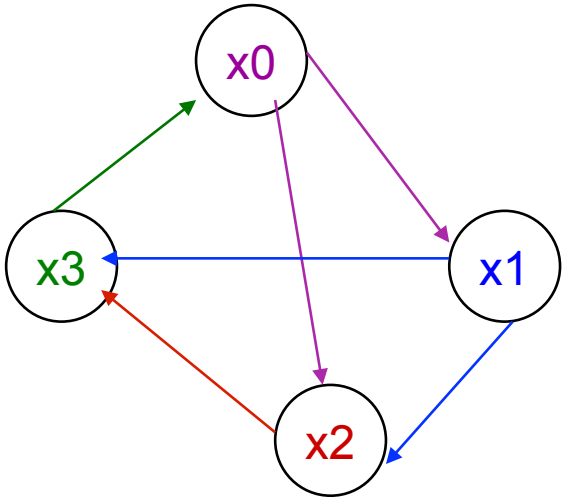


1.sends data to n2
2.Calculates Cov_{n0n0}
3.Calculates the remaining components of Cov $_{n0n2}$

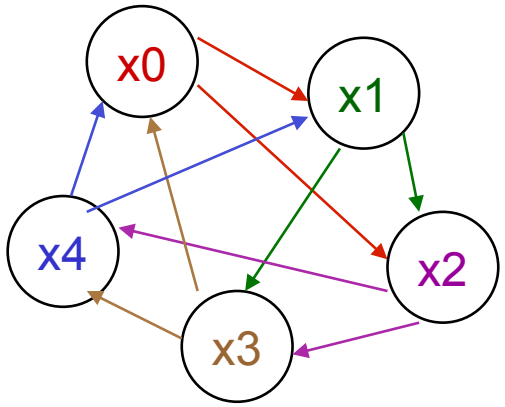
1.sends data to n2
2.Calculates Cov_{n1n1}
3.Calculates Cov $_{n0n1}$


Cov n0n0
Cov n1n1
xx xy xz xw xl
yx yy yz yw yl
zx zy zz zw zl
wx wy wz ww wl
lx ly lz lw ll
Cov n2n2

Generalization with n nodes



If the total number of nodes is even
i.e. t = 2r ,where r>=1
i)send Xj ,where j=0 to r-1 to its r successive nodes
ii)send Xj ,where j=r to 2r-1 to its r-1 successive nodes
iii)Compute Cv(Xj,k) parallel y at Sk



if the total number of sites/nodes is odd
i.e. t= 2r+1 ,where r>=1
i)send Xj,where j=0 to 2r to its r successive nodes
ii)Compute Cv(Xj,k) parallel y at Sk

FeedBack !!