

Mixed metrics

Comparing heterogeneous objects

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Ascendant hierarchical clustering

Hot points :

➤ Agregation criteria

- Centroid criterion
- Ward's criterion

➤ Distance between individuals

- Only quantitative variables → Euclidean, Correlations
- Only qualitative variables → χ^2 (Benzécri 80)
- Heterogenous variables → Gower (71)
(compatibility measures)
 - Gowda i Diday (91)
 - Gibert's Mixed (91)
 - Ichino i Yaguchi (94)
 - Ralambondrainy (95)
 - Generalized Gibert's Mixed (13)

Ascendant hierarchical clustering

- Benzécri, J. P. (1980). *Pratique de l'analyse des données*. vol. 1, analyse des correspondances.
- J. C. Gower **A General Coefficient of Similarity and Some of Its Properties** *Biometrics*, Vol. 27, No. 4. (Dec., 1971), pp. 857-871
- Diday, E. and Gowda, K.C. Symbolic clustering using a new similarity measure *IEEE Trans. on systems, mans, and cybernetics* 22(2) 368-378, 1992
- Gibert, K. (1991). *Klass. Estudi d'un sistema d'ajuda al tractament estadístic de grans bases de dades* (Doctoral dissertation, Master's thesis, UPC).
- Gibert, K., and Cortés, U. (1997). "Weighing quantitative and qualitative variables in clustering methods." *Mathware and Soft Computing*, 4(3), 251-266
- Ichino, M. and Yaguchi, H. (1990), Generalized Minkowski metrics for mixed features. *Electron. Comm. Jpn. Pt. III*, 73: 12–20. doi: 10.1002/ecjc.4430730602
- [Ralambondrainy 1995] Ralambondrainy, H A conceptual version of the K-means algorithm *Lifetime Learning Publications*, Belmont, California, 1995
- Gibert K, Valls A, Batet M. (2014): Introducing semantic variables in mixed distance measures. Impact on hierarchical clustering, *Knowledge and Information Systems* 40(3):559-593, Springer, DOI: 10.1007/s10115-013-0663-5.

Proximity measures

Distance:

$$d : E \times E \longrightarrow \mathbb{R}^+$$

$$1. \forall x \in E \quad d(x, x) = 0$$

$$2. \forall x, \forall y \quad d(x, y) = d(y, x)$$

$$3. \forall x, \forall y, \forall z \quad d(x, z) \leq d(x, y) + d(y, z)$$

Distance index:

Verifies 1. and 2., but 3.

Similarity index:

$$s : E \times E \longrightarrow \mathbb{R}^+$$

$$1. \forall x, \forall y \quad s(x, y) = s(y, x)$$

$$2. \forall x, \forall y \quad s(x, x) = s(y, y) > s(y, z)$$

Dissimilarity index:

$$1. \forall x, \forall y \quad d_s(x, y) = d_s(y, x)$$

$$2. \forall x, \forall y \quad d_s(x, x) = d_s(y, y) < d_s(y, z)$$

$$d_s(x, y) = s(x, x) - s(x, y)$$

Euclidean

$$d^2(i, i') = \sum_{j=1}^p (x_{ij} - x_{i'j})^2$$

Distances

■ **Minkowski distances**: $d(i, i') = \sum_{j=1}^p (|x_{ij} - x_{i'j}|^d)^{1/d}$

– Euclidean (L_2):

$$d^2(i, i') = \sum_{j=1}^p (x_{ij} - x_{i'j})^2$$

– Manhattan (L_1) (absolute value):

$$d(i, i') = \sum_{j=1}^p |x_{ij} - x_{i'j}|$$

– Chebichev (L_∞):

$$d(i, i') = \max_j |x_{ij} - x_{i'j}|$$

– Correlation

$$d^2(x_j, x_{j'}) = 2(1 - \text{cor}(x_j, x_{j'}))$$

(L_2 with standardized *variables*)

– Cosine similarity between individuals $1 - \cos(x_i, x_{i'}) = 1 - \frac{\langle x_i, x_{i'} \rangle}{\|x_i\| \|x_{i'}\|}$

Distances

- To scale or not to scale
 - Standardization

$$x_{ij} \leftarrow \frac{x_{ij} - \bar{x}_j}{s_j}$$

Better to
Scale

- Max normalization

$$x_{ij} \leftarrow \frac{x_{ij}}{\max(x_j) - \min(x_j)}$$

Context
required to
scale

Distance correlation

Pearson correlation:

For linear association
between two variables

$$\text{cov}(x, y) = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})(y_i - \bar{y})$$

$$r_{xy} = \text{cor}(x, y) = \frac{\text{cov}(x, y)}{\sqrt{\text{cov}(x, x) \cdot \text{cov}(y, y)}}$$

Distance correlation:

For non linear association
between two sets of variables
(eventually two variables)

(x, y) Let x and y be two random vectors of p and q dimensions

$a_{kl} = d(x_k, x_l) \quad k, l = 1, \dots, n$ Let a and b all pair distances between
the x and y vectors respectively

$b_{kl} = d(y_k, y_l) \quad k, l = 1, \dots, n$

$\bar{\bar{a}}_{kl} = a_{kl} - a_{k.} - a_{.l} + a_{..}$ (double centering of the distance matrices)

$\bar{\bar{b}}_{kl} = b_{kl} - b_{k.} - b_{.l} + b_{..}$

Distance covariance

$$dv^2(x, y) = \frac{1}{n^2} \sum_{k=1}^n \sum_{l=1}^n \bar{\bar{a}}_{kl} \times \bar{\bar{b}}_{kl}$$

Distance standard deviation

$$dv(x, x)$$

Distance correlation

$$dc^2(x, y) = \frac{dv^2(x, y)}{dv(x, x) \times dv(y, y)}$$

Distance

Binary variables (Logical)

$$d(i, i') = f(n_{11}, n_{00}, n_{10}, n_{01})$$

	<i>Object i</i>		
	n_{11}	n_{10}	
<i>Object i'</i>	n_{01}	n_{00}	← Uninformative

Jaccard

$$d(i, i') = \frac{n_{10} + n_{01}}{n_{10} + n_{01} + n_{11}}$$

Assymetrical distance

Sokal

$$d(i, i') = \frac{n_{10} + n_{01}}{n}$$

Symetrical distance

Distances

Nominal variables

Simple matching coefficient (Hamming)

$$d(i, i') = \frac{\# \text{ vars mismatch}}{p}$$

Chi-2 distance

$$X^2 = \sum_{k=1}^p \sum_{j=1}^q \frac{\left(n_{kj} - \frac{n_k n_j}{n} \right)^2}{\frac{n_k n_j}{n}}$$

Distances

Ordinal variables

normalized ranks

$$x_{ij} = \frac{r_{ij} - 1}{R_j - 1} \in [0, 1]$$

treat them as numerical

Distances

strings and curves

■ Edit distance between two strings (or DNA sequences)

Minimum number of operations to transform one string into the other (removals, insertions and substitutions). It verifies the three axioms of a distance. Operations can be weighted inversely to their likelihood.

What is the edit distance between "SEAL" and "ATE"?

Step	Comparison	Edit necessary	Total Editing
1.	"A" to "S"	replace: +1 edit	"STE" in 1 edit
2.	"T"	delete: +1 edit	"SE" in 2 edits
3.	"A"	insertion: +1 edit	"SEA" in 3 edits
4.	"L"	insertion: +1 edit	"SEAL" in 4 edits

■ Hamming distance between two strings of the same length

Counts the minimum number of substitutions to convert one string in the other

$\text{dist}(\text{cats}, \text{dogs}) = 3/4$

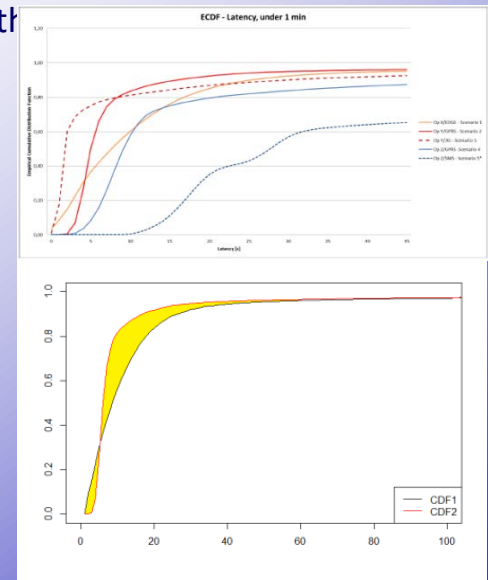
cats => dats (substitute 'd' for 'c')

dats => dots (substitute 'o' for 'a')

dots => dogs (substitute 'g' for 't')

■ Vertical distance between curves $\max |F(x) - G(x)|$

■ Histogram matching distance $\int |F(x) - G(x)|$



Management of heterogeneous data matrices

Heterogeneous matrices faced. Several approaches [Anderberg 73]:

Variables partitioning

(Ex, naive bayes classifier)

Variables converting

Compatibility measures

Mixed metrics required

Gower

[Gow 71]

$$s(i, i') = \frac{\sum_{k=1}^K w_k(i, i') s_k(i, i')}{\sum_{k=1}^K w_k(i, i')} ;$$

$$d(i, i') = 1 - s(i, i')$$

$$w_k(i, i') = \begin{cases} 0 & \text{if } (x_{ik} = \text{missing}) \text{ or } (x_{i'k} = \text{missing}) \\ 0 & \text{if } (X_k \text{ binary}) \text{ and } (x_{ik} = \text{false}) \text{ i } (x_{i'k} = \text{false}) \\ & \text{and negative absence of } X_k \text{ excluded} \\ 1 & \text{otherwise} \end{cases}$$

$$s_k(i, i') = \begin{cases} 1 - \frac{|x_{ik} - x_{i'k}|}{R_k} & \text{if } X_k \text{ numerical} \\ 1 & \text{if } (X_k \text{ qualitative}) \text{ and } (x_{ik} = x_{i'k}) \\ 0 & \text{if } (X_k \text{ qualitative}) \text{ and } (x_{ik} \neq x_{i'k}) \end{cases}$$

Gowda-Diday

[Gow 91]

$$D(i, i') = \sum_{k=1}^K D_k(i_k, i'_k) \quad \text{with} \quad D_k(i, i') = D_p(i, i') + D_s(i, i') + D_c(i, i')$$

Component Position

$$D_{kp}(i, i') = \begin{cases} \frac{|x_{ik} - x_{i'k}|}{R_k} & \text{if } (X_k \text{ numerical}) \text{ and } R_k \text{ is rang of } X_k \\ 0 & \text{if } (X_k \text{ qualitative}) \end{cases}$$

Component Span

$$D_{ks}(i, i') = \begin{cases} 0 & \text{if } X_k \text{ numerical} \\ 0 & \text{if } X_k \text{ qualitative and not multivalued} \end{cases}$$

Component Content

$$D_{kc}(i, i') = \begin{cases} 0 & \text{if } X_k \text{ numerical} \\ 0 & \text{if } (X_k \text{ qualitative}) \text{ and } (x_{ik} = x_{i'k}) \\ 1 & \text{if } (X_k \text{ qualitative}) \text{ and } (x_{ik} \neq x_{i'k}) \end{cases}$$

Mixed Metrics

[Gib 91]

$$d^2_{(\alpha,\beta)}(i, i') = \alpha d_\zeta^2(i, i') + \beta d_Q^2(i, i')$$

$$d_\zeta^2(i, i') = \sum_{\forall k \in \zeta} \frac{(x_{ik} - x_{i'k})^2}{s_k^2}$$

$$d_k^2(i, i') = \begin{cases} 0, & \text{if } x_{ik} = x_{i'k} \\ \frac{1}{I_k^i} + \frac{1}{I_k^{i'}}, & \text{otherwise, for compact } i \text{ and } i' \text{ with respect } X_k \\ \frac{(f_i^{k_s} - 1)^2}{I^{k_s}} + \sum_{j \neq s} \frac{(f_i^{k_j})^2}{I^{k_j}}, & \text{if } x_{ik} = c_s^k, \text{ and extended } i' \text{ with respect } X_k \\ \sum_{j=1}^{n_k} \frac{(f_i^{k_j} - f_{i'}^{k_j})^2}{I^{k_j}}, & \text{for } i, i' \text{ extended with respect } X_k \end{cases}$$

if $x_{ik} = x_{i'k}$

otherwise, for compact i and i' with respect X_k

if $x_{ik} = c_s^k$, and extended i' with respect X_k

for i, i' extended with respect X_k

■ **Proposal [Gib 91]:**

$$\alpha = \frac{n_\zeta}{d_\zeta^2 \max^*}$$

$$\beta = \frac{n_Q}{d_Q^2 \max^*}$$

Ralambondrainy

[Ra195]

$$d^2(i, i') = \pi_1 d_{1/\sigma^2}^2(i, i') + \pi_2 d_{\chi^2}^2(i, i')$$

■ Proposal [Ra1 88] :

■ Standardisation by the inertia

$$\pi_1 = \frac{1}{\text{Card}(\zeta)}$$

$$\pi_2 = \frac{1}{n_k - 1}$$

■ Standardisation by the norm

$$\pi_1 = \frac{1}{\sqrt{\sum \{\rho^2(X_k, X_{k'}) / k, k' \in \zeta\}}}$$

$$\pi_2 = \sqrt{n_k - 1}$$

Ichino-Yaguchi

[Ichi 94]

Generalized Minkowski metrics,
p-order ($p \geq 1$)

$$d_p(i, i') = \sqrt[p]{\sum_{k=1}^K \left(\frac{\phi(x_{ik}, x_{i'k})}{|U_k|} \right)^p}$$

where $|U_k|$ Normalizes (with R_k o n_k)

$\phi(x_{ik}, x_{i'k})$ Is a function of :

• Cartesian Joint

$$|x_{ik} \oplus x_{i'k}| = \begin{cases} |x_{ik} - x_{i'k}|, & \text{if } X_k \text{ numerical} \\ 1, & \text{if } X_k \text{ categorical and } x_{ik} = x_{i'k} \\ 2, & \text{if } X_k \text{ categorical and } x_{ik} \neq x_{i'k} \end{cases}$$

• Catesian Meet

$$|x_{ik} \oplus x_{i'k}| = \begin{cases} |x_{ik} - x_{i'k}|, & \text{if } X_k \text{ numerical} \\ 1, & \text{if } X_k \text{ categorical and } x_{ik} = x_{i'k} \\ 2, & \text{if } X_k \text{ categorical and } x_{ik} \neq x_{i'k} \end{cases}$$

• Cardinality of $x, |x_{ik}|, (0 \text{ o } 1)$

• Factor $\gamma \in [0, 0.5]$

Example Michalski data

(Gibert mixed metrics)

	A	B	C	F	J	M	P	R	S	T
A	0	1.2488811	1.2438452	0.8309088	1.1683081	1.2438452	0.8309088	0.63954794	1.4049911	0.90644586
B	1.2488811	0	0.8309088	1.2438452	0.90644586	1.4352059	0.63954794	0.8309088	0.8309088	1.1683081
C	1.2438452	0.8309088	0	1.3999553	1.1330574	1.0474486	1.6920323	1.0474486	1.8229634	0.7201209
F	0.8309088	1.2438452	1.3999553	0	0.7201209	1.2388093	1.5006716	1.2388093	1.2488811	1.1330574
J	1.1683081	0.90644586	1.1330574	0.7201209	0	0.97191143	1.1632721	1.5762087	1.2942033	1.2488811
M	1.2438452	1.4352059	1.0474486	1.2388093	0.97191143	0	1.2488811	1.2085946	2.2661147	1.1632721
P	0.8309088	0.63954794	1.6920323	1.5006716	1.1632721	1.2488811	0	1.2488811	0.63451207	0.97191143
R	0.63954794	0.8309088	1.0474486	1.2388093	1.5762087	1.2085946	1.2488811	0	2.2661147	1.7675694
S	1.4049911	0.8309088	1.8229634	1.2488811	1.2942033	2.2661147	0.63451207	2.2661147	0	0.7201209
T	0.90644586	1.1683081	0.7201209	1.1330574	1.2488811	1.1632721	0.97191143	1.7675694	0.7201209	0

Ralambondrainy normalized by inertia

$\pi_1=0.5$, $\pi_2=0.2$

	A	B	C	F	J	M	P	R	S	T
A	0	1.2560605	0.84210527	1.7314991	0.8254386	0.84210527	1.7314991	0.70877194	4.7996807	1.7481657
B	1.2560605	0	1.7314991	0.84210527	1.7481657	1.8648324	0.70877194	1.7314991	1.7314991	0.8254386
C	0.84210527	1.7314991	0	1.2893938	0.25	2.5017545	3.6244814	2.5017545	4.3242416	1.1393938
F	1.7314991	0.84210527	1.2893938	0	1.1393938	3.5244815	2.6017544	3.5244815	1.2560605	0.25
J	0.8254386	1.7481657	0.25	1.1393938	0	2.4850879	3.507815	2.618421	4.2075753	1.2560605
M	0.84210527	1.8648324	2.5017545	3.5244815	2.4850879	0	1.2560605	0.26666668	6.692663	3.507815
P	1.7314991	0.70877194	3.6244814	2.6017544	3.507815	1.2560605	0	1.2560605	3.391148	2.4850879
R	0.70877194	1.7314991	2.5017545	3.5244815	2.618421	0.26666668	1.2560605	0	6.692663	3.641148
S	4.7996807	1.7314991	4.3242416	1.2560605	4.2075753	6.692663	3.391148	6.692663	0	1.1393938
T	1.7481657	0.8254386	1.1393938	0.25	1.2560605	3.507815	2.4850879	3.641148	1.1393938	0

Ralambondrainy normalized by norm, pi1= 0,617, p2=2,236

	A	B	C	F	J	M	P	R	S	T
A	0	3.8718193	3.5263379	3.298699	3.3399992	3.5263379	3.298699	2.035626	7.0879183	3.4850378
B	3.8718193	0	3.298699	3.5263379	3.4850378	4.7894106	2.035626	3.298699	3.298699	3.3399992
C	3.5263379	3.298699	0	4.2444973	2.795085	4.415724	6.796831	4.415724	7.6610384	2.5674462
F	3.298699	3.5263379	4.2444973	0	2.5674462	5.678797	5.533758	5.678797	3.8718193	2.795085
J	3.3399992	3.4850378	2.795085	2.5674462	0	4.229385	5.492458	5.7200966	6.356665	3.8718193
M	3.5263379	4.7894106	4.415724	5.678797	4.229385	0	3.8718193	2.981424	10.58605	5.492458
P	3.298699	2.035626	6.796831	5.533758	5.492458	3.8718193	0	3.8718193	4.1880846	4.229385
R	2.035626	3.298699	4.415724	5.678797	5.7200966	2.981424	3.8718193	0	10.58605	6.9831696
S	7.0879183	3.298699	7.6610384	3.8718193	6.356665	10.58605	4.1880846	10.58605	0	2.5674462
T	3.4850378	3.3399992	2.5674462	2.795085	3.8718193	5.492458	4.229385	6.9831696	2.5674462	0

Gower

	A	B	C	F	J	M	P	R	S	T
A	0	0.625	0.625	0.5	0.625	0.625	0.5	0.375	0.625	0.5
B	0.625	0	0.5	0.625	0.5	0.75	0.375	0.5	0.5	0.625
C	0.625	0.5	0	0.625	0.5	0.5	0.875	0.5	0.75	0.375
F	0.5	0.625	0.625	0	0.375	0.625	0.75	0.625	0.625	0.5
J	0.625	0.5	0.5	0.375	0	0.5	0.625	0.75	0.5	0.625
M	0.625	0.75	0.5	0.625	0.5	0	0.625	0.5	1.0	0.625
P	0.5	0.375	0.875	0.75	0.625	0.625	0	0.625	0.375	0.5
R	0.375	0.5	0.5	0.625	0.75	0.5	0.625	0	1.0	0.875
S	0.625	0.5	0.75	0.625	0.5	1.0	0.375	1.0	0	0.375
T	0.5	0.625	0.375	0.5	0.625	0.625	0.5	0.875	0.375	0

Example

Michalski data

Gower

	A	B	C	F	J	M	P	R	S	T
A	0	0.625	0.625	0.5	0.625	0.625	0.5	0.375	0.625	0.5
B	0.625	0	0.5	0.625	0.5	0.75	0.375	0.5	0.5	0.625
C	0.625	0.5	0	0.625	0.5	0.5	0.875	0.5	0.75	0.375
F	0.5	0.625	0.625	0	0.375	0.625	0.75	0.625	0.625	0.5
J	0.625	0.5	0.5	0.375	0	0.5	0.625	0.75	0.5	0.625
M	0.625	0.75	0.5	0.625	0.5	0	0.625	0.5	1.0	0.625
P	0.5	0.375	0.875	0.75	0.625	0.625	0	0.625	0.375	0.5
R	0.375	0.5	0.5	0.625	0.75	0.5	0.625	0	1.0	0.875
S	0.625	0.5	0.75	0.625	0.5	1.0	0.375	1.0	0	0.375
T	0.5	0.625	0.375	0.5	0.625	0.625	0.5	0.875	0.375	0

Gibert mixed metrics

	A	B	C	F	J	M	P	R	S	T
A	0	1.2488811	1.2438452	0.8309088	1.1683081	1.2438452	0.8309088	0.63954794	1.4049911	0.90644586
B	1.2488811	0	0.8309088	1.2438452	0.90644586	1.4352059	0.63954794	0.8309088	0.8309088	1.1683081
C	1.2438452	0.8309088	0	1.3999553	1.1330574	1.0474486	1.6920323	1.0474486	1.8229634	0.7201209
F	0.8309088	1.2438452	1.3999553	0	0.7201209	1.2388093	1.5006716	1.2388093	1.2488811	1.1330574
J	1.1683081	0.90644586	1.1330574	0.7201209	0	0.97191143	1.1632721	1.5762087	1.2942033	1.2488811
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T	0.90644586	1.1683081	0.7201209	1.1330574	1.2488811	1.1632721	0.97191143	1.7675694	0.7201209	0

Ralambondrainy, Pi1=0,1, Pi2=0,2

	A	B	C	F	J	M	P	R	S	T
A	0	1.2560605	0.84210527	1.7314991	0.8254386	0.84210527	1.7314991	0.70877194	4.7996807	1.7481657
B	1.2560605	0	1.7314991	0.84210527	1.7481657	1.8648324	0.70877194	1.7314991	1.7314991	0.8254386
C	0.84210527	1.7314991	0	1.2893938	0.25	2.5017545	3.6244814	2.5017545	4.3242416	1.1393938
F	1.7314991	0.84210527	1.2893938	0	1.1393938	3.5244815	2.6017544	3.5244815	1.2560605	0.25
J	0.8254386	1.7481657	0.25	1.1393938	0	2.4850879	3.507815	2.618421	4.2075753	1.2560605
M	0.84210527	1.8648324	2.5017545	3.5244815	2.4850879	0	1.2560605	0.26666668	6.692663	3.507815
P	1.7314991	0.70877194	3.6244814	2.6017544	3.507815	1.2560605	0	1.2560605	3.391148	2.4850879
R	0.70877194	1.7314991	2.5017545	3.5244815	2.618421	0.26666668	1.2560605	0	6.692663	3.641148
S	4.7996807	1.7314991	4.3242416	1.2560605	4.2075753	6.692663	3.391148	6.692663	0	1.1393938
T	1.7481657	0.8254386	1.1393938	0.25	1.2560605	3.507815	2.4850879	3.641148	1.1393938	0

Ralambondrainy, Pi1=0,617, Pi2=0,2236

	A	B	C	F	J	M	P	R	S	T
A	0	3.8718193	3.5263379	3.298699	3.3399992	3.5263379	3.298699	2.035626	7.0879183	3.4850378
B	3.8718193	0	3.298699	3.5263379	3.4850378	4.7894106	2.035626	3.298699	3.298699	3.3399992
C	3.5263379	3.298699	0	4.2444973	2.795085	4.415724	6.796831	4.415724	7.6610384	2.5674462
F	3.298699	3.5263379	4.2444973	0	2.5674462	5.678797	5.533758	5.678797	3.8718193	2.795085
J	3.3399992	3.4850378	2.795085	2.5674462	0	4.229385	5.492458	5.7200966	6.356665	3.8718193
M	3.5263379	4.7894106	4.415724	5.678797	4.229385	0	3.8718193	2.981424	10.58605	5.492458
P	3.298699	2.035626	6.796831	5.533758	5.492458	3.8718193	0	3.8718193	4.1880846	4.229385
R	2.035626	3.298699	4.415724	5.678797	5.7200966	2.981424	3.8718193	0	10.58605	6.9831696
S	7.0879183	3.298699	7.6610384	3.8718193	6.356665	10.58605	4.1880846	10.58605	0	2.5674462
T	3.4850378	3.3399992	2.5674462	2.795085	3.8718193	5.492458	4.229385	6.9831696	2.5674462	0