Persistent homology and Mayer-Vietoris formulas

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Shape recognition











- It is the task of finding a given object in a scene/dataset/ image/video sequence.
- Humans recognize a multitude of objects in images with little effort, despite variations in
 - different view points,
 - different sizes / scale
 - o translations and rotations
 - o partial obstructions from view.
- Increasing interest in automatic shape recognition
- This task is still a challenge for computer vision systems in general.
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What is the shape of an object?



- Tentative definitions are generally based on *observers' perceptions*.
- Dependence on observers implies large subjectivity
 - o changes due to object orientation and distance from the object
 - changes due to light conditions
- Human judgments focus on persistent perceptions
 - Non-persistent properties can be considered as noise.
 - Only *stable perceptions* concur to give a shape to objects.





Similar shapes with respect to what?



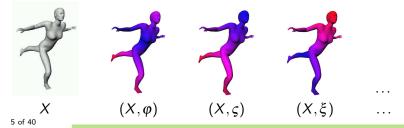
How to model observations and perceptions?



- An observation can be modeled as a topological space X.
 - o it depends on what the observer is observing.
- Observer's perceptions can be modeled as a function $f: X \to \mathbb{R}^n$.
 - o it depends on the shape property the observer is perceiving.

Thus to model a shape we consider pairs (X, f) where

- X is a topological space
- $f: X \to \mathbb{R}$ is a (continuous) function.



Comparing shapes



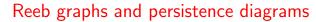
• How can we compare two pairs (X, φ) , (Y, ψ) ?

$$d\left(\begin{array}{c} \\ (X,\varphi) \\ \end{array}, \begin{array}{c} \\ (Y,\psi) \end{array}\right) = ?$$

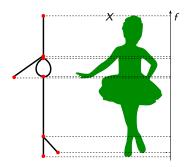
 Persistence allows us to describe such a pair by means of suitable shape descriptors (persistence diagrams).



Instead of comparing shapes, we can compare shape descriptors.

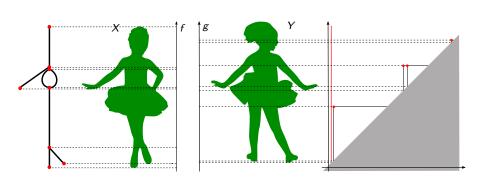






Reeb graphs and persistence diagrams





Size functions



- For every $u \in \mathbb{R}$, let us denote by X_u the **lower level set** $\{p \in X : f(p) \le u\}$.
- Let Δ^+ be the open half plane $\{(u,v) \in \mathbb{R}^2 : u < v\}$.

Size functions

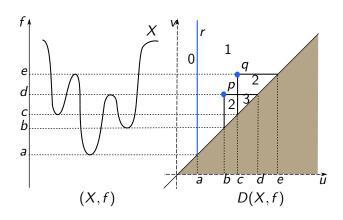


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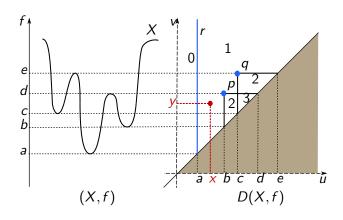
The **size function** associated with the pair (X, f) is the function $D(X, f): \Delta^+ \to \mathbb{N}$ that takes each $(u, v) \in \Delta^+$ to the number of connected components in X_v containing at least one point of X_u .

$$D(X, f)(u, v) = \dim H_0^{u,v}(X) = \dim \operatorname{im}(H_0(X_u) \to H_0(X_v)).$$

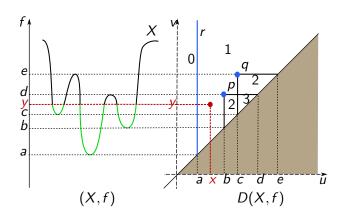




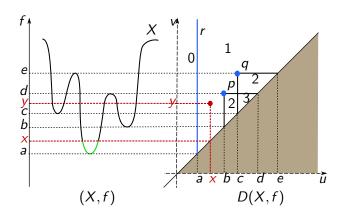






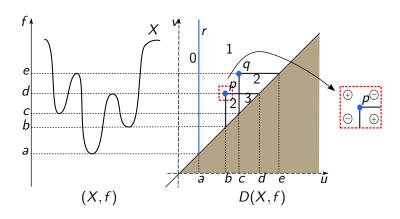






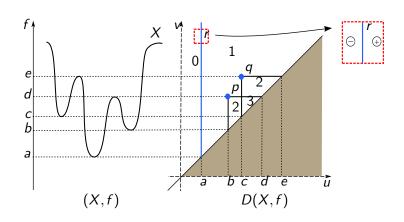






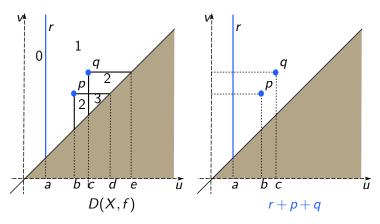
Points of a persistence diagram





Points of a persistence diagram

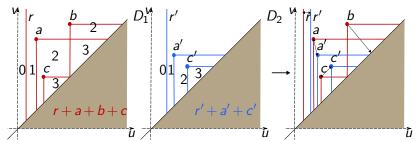




D(X, f) is equivalent to r + p + q.

Hausdorff distance





Let D_1 and D_2 be two persistence diagrams with the same number of points at infinity. Let A_1 (resp. A_2) be the set of all points for D_1 (resp. D_2), augmented by adding a countable infinity of points of the diagonal Δ . The **Hausdorff distance** between D_1 and D_2 is given by

$$d_H(D_1, D_2) = \max \{ \max_{p \in A_1} \min_{q \in A_2} \|p - q\|_{\infty}, \max_{q \in A_2} \min_{p \in A_1} \|q - p\|_{\infty} \}.$$

Persistent homology groups



• For every $u \in \mathbb{R}$, let us denote by X_u the **lower level set** $\{p \in X : f(p) \le u\}$.

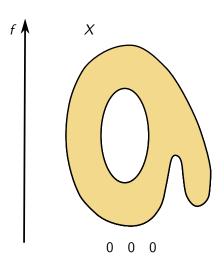
Persistent homology groups



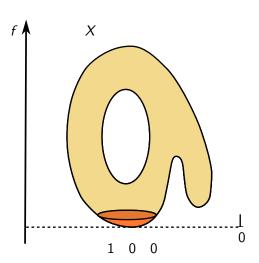
• For every $u \in \mathbb{R}$, let us denote by X_u the **lower level set** $\{p \in X : f(p) \le u\}$.

Given a pair (X, φ) , and $u, v \in \mathbb{R}$, with u < v, we shall denote by $\iota^{u,v}$ the inclusion of X_u into X_v . This mapping induces a homomorphism of homology groups $\iota^{u,v}_k : H_k(X_u) \to H_k(X_v)$ for each integer $k \ge 0$. The kth persistent homology group $H^{u,v}_k(X,f)$ is the image of the homomorphism $\iota^{u,v}_k : H_k(X_u) \to H_k(X_v)$, that is $H^{u,v}_k(X,f) = \operatorname{im} \iota^{u,v}_k$.

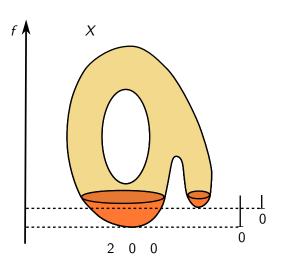




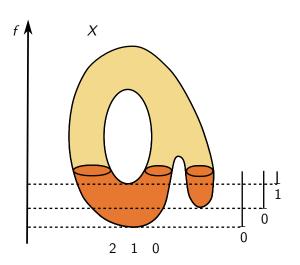




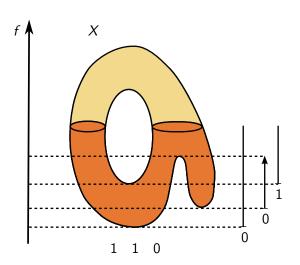




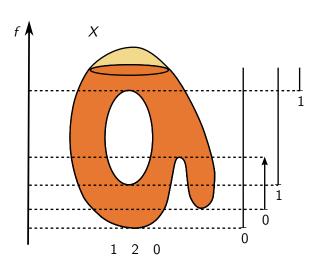




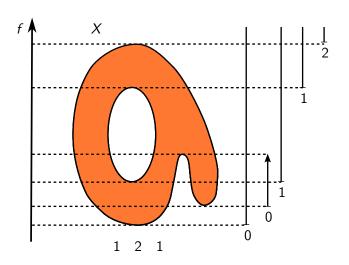








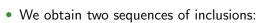


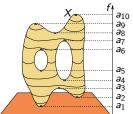


Sub- and super-level set filtrations



- Let X be a triangulable subspace of \mathbb{R}^n .
- Let $f: X \to \mathbb{R}$ be a continuous function with a finite number of homological critical values $a_1 < a_2 < ... < a_r$.
- For $s_0 < s_1 < \ldots < s_r$ such that $s_{i-1} < a_i < s_i$ we have
 - sub-level sets $X_i = f^{-1}((-\infty, s_i]),$
 - super-level sets $X^i = f^{-1}([s_i, +\infty)).$





$$\emptyset = X_0 \hookrightarrow \ldots \hookrightarrow X_i \hookrightarrow \ldots \hookrightarrow X_j \hookrightarrow \ldots X_r = X,$$

$$(X,\emptyset) = (X,X^r) \hookrightarrow \ldots \hookrightarrow (X,X^j) \hookrightarrow \ldots \hookrightarrow (X,X^i) \hookrightarrow \ldots (X,X^0) = (X,X).$$

Persistent homology



• Apply the homology functor:

$$\begin{array}{ccc} H_k(X_0) \rightarrow \dots \rightarrow H_k(X_i) \rightarrow \dots \rightarrow H_k(X_j) \rightarrow \dots \rightarrow H_k(X) & 0 \\ & & & & || & & || \\ 0 & & & H_k(X,X^r) \rightarrow \dots \rightarrow H_k(X,X^j) \rightarrow \dots \rightarrow H_k(X,X^i) \rightarrow \dots \rightarrow H_k(X,X^0) \end{array}$$

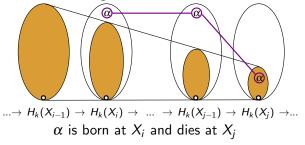
Persistent homology



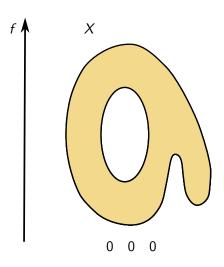
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$$\begin{array}{ccc} H_k(X_0) \rightarrow \dots \rightarrow H_k(X_i) \rightarrow \dots \rightarrow H_k(X_j) \rightarrow \dots \rightarrow H_k(X) & 0 \\ & & & & & || & & || \\ 0 & & & & H_k(X,X^r) \rightarrow \dots \rightarrow H_k(X,X^j) \rightarrow \dots \rightarrow H_k(X,X^i) \rightarrow \dots \rightarrow H_k(X,X^0) \end{array}$$

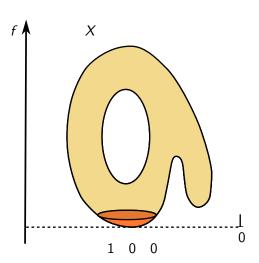
 analyze the scale at which a homological feature is created, and when it is annihilated along the filtration:



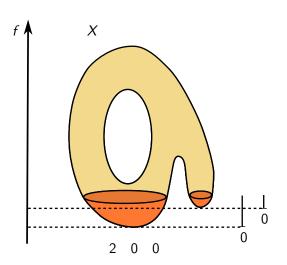




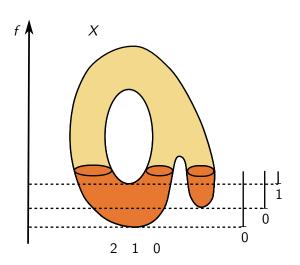




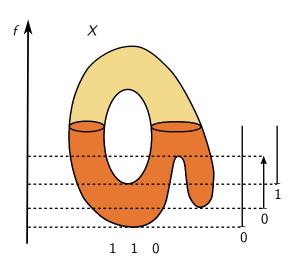




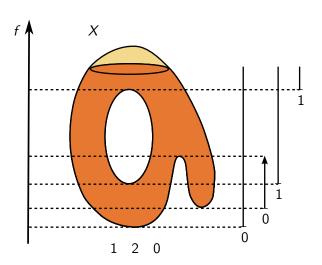






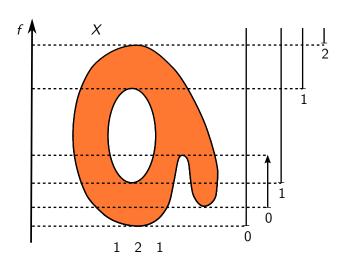




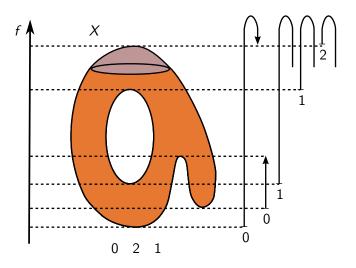


Ordinary persistence

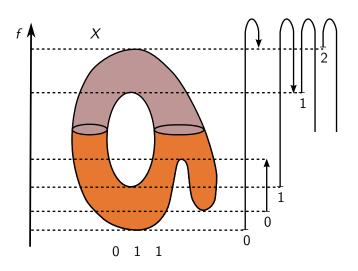




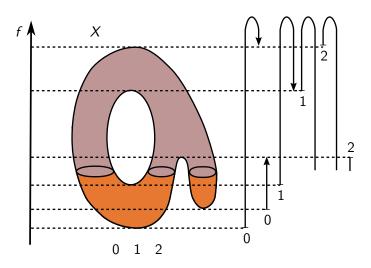




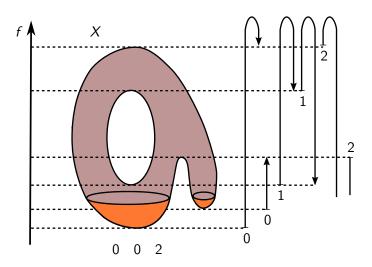




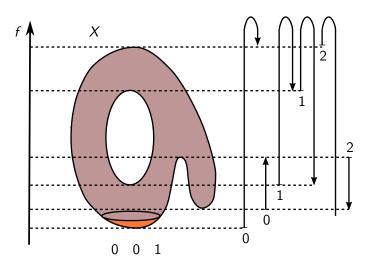




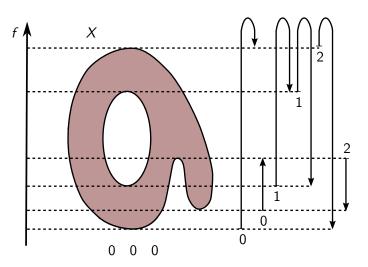












Persistent homology



Define the ordinary, relative ed extended persistent Betti numbers as follows:

$$H_{k}(X_{0}) \rightarrow \dots \rightarrow H_{k}(X_{i}) \rightarrow \dots \rightarrow H_{k}(X_{j}) \rightarrow \dots \rightarrow H_{k}(X)$$

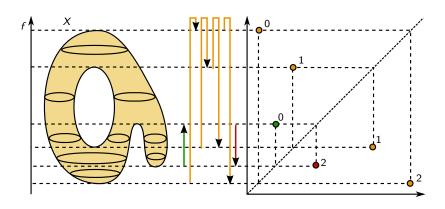
$$0$$

$$H_{k}(X,X^{r}) \rightarrow \dots \rightarrow H_{k}(X,X^{j}) \rightarrow \dots \rightarrow H_{k}(X,X^{j}) \rightarrow \dots \rightarrow H_{k}(X,X^{0})$$

- $Ord_k^{i,j}(X) := \operatorname{rk} H_k(X_i \hookrightarrow X_j)$, per i < j
- $Rel_k^{j,i}(X) := \operatorname{rk} H_k((X,X^j) \hookrightarrow (X,X^i))$, per $i \leq j$
- $Ext_k^{i,j}(X) := rkH_k(X_i \hookrightarrow (X,X^j)).$

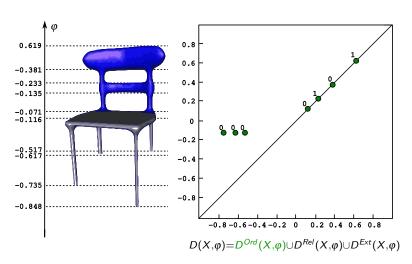


Persistence diagrams $D^{Ord}(X, f), D^{Rel}(X, f), D^{Ext}(X, f)$



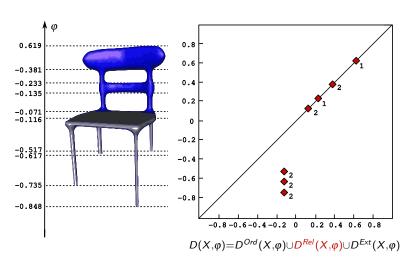
Persistence diagrams





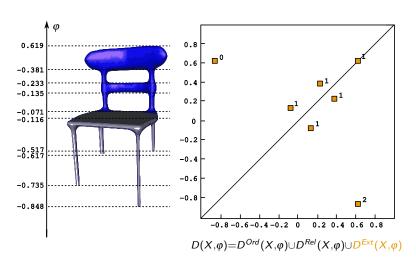
Persistence diagrams





Persistence diagrams





Shape recognition from partial information



It is the task of detecting sub-parts similarities between objects possibly having different overall shape.

• Recognition/classification of 2D content has to handle occlusions









 3D content is based on the full representation of the shape, nevertheless recognition of partial similarities may help in the handling of non-rigid and articulated objects.





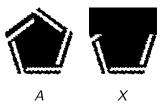




Recognition under occlusion: how do we model the problem?

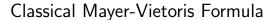
blem?

- A is the object of interest;
- B the occluding pattern;
- $X = A \cup B$ the occluded object;
- $f: X \to \mathbb{R}$ the measuring function.



Assuming X, A, B compact, locally connected Hausdorff spaces, what is the relation among

$$D(X,f), D(A,f|_A), D(B,f|_B)$$
?





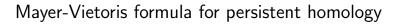
Given a triad (X, A, B) with $X = A \cup B$, a Mayer-Vietoris formula is a relation between the Betti numbers of X, A, B, and $C = A \cap B$:

$$\operatorname{rk} H_k(X) = \operatorname{rk} H_k(A) + \operatorname{rk} H_k(B) - \operatorname{rk} H_k(C) + \operatorname{rk} \delta_k + \operatorname{rk} \delta_{k-1}.$$

It is obtained from the Mayer-Vietoris sequence

$$\cdots \to H_{k+1}(X) \stackrel{\delta_k}{\to} H_k(C) \to H_k(A) \oplus H_k(B) \to H_k(X) \stackrel{\delta_{k-1}}{\to} \cdots,$$

when this is exact.





We use

- Čech homology
- · homology coefficients in a vector space

and assume that

- $X = A \cup B$, A, B, $C = A \cap B$ are triangulable subspaces of some \mathbb{R}^n ,
- the homology groups of the sub- and super-level sets of f, $f|_A$, $f|_B$, $f|_C$ are finitely generated.

Mayer-Vietoris formula for ordinary persistence



• For every $u < v \in \mathbb{R}$, and $k \in \mathbb{Z}$

$$Ord_k^{u,v}(X) = Ord_k^{u,v}(A) + Ord_k^{u,v}(B) - Ord_k^{u,v}(C) + \operatorname{rk}\delta_k' - \operatorname{rk}\delta_k'' + \operatorname{rk}\delta_{k-1}$$

with δ'_k , δ''_k , and δ_{k-1} as in the following diagram:

Mayer-Vietoris formula for relative persistence



• For every $v \le u \in \mathbb{R}$, and $k \in \mathbb{Z}$

with
$$\bar{\delta}_k'$$
, $\bar{\delta}_{k-1}''$, and $\bar{\delta}_{k-1}$ as in the following diagram:

 $H_{k+1}(X,X^u) \stackrel{\bar{\delta}_k}{\to} H_k(C,C^u) \to H_k(A,A^u) \oplus H_k(B,B^u) \to H_k(X,X^u)$
 $\downarrow \bar{h}_{k+1} \downarrow \bar{f}_k \downarrow \bar{g}_k \downarrow \bar{h}_k$
 $H_{k+1}(X,X^v) \stackrel{\bar{\delta}_k'}{\to} H_k(C,C^v) \to H_k(A,A^v) \oplus H_k(B,B^v) \to H_k(X,X^v)$
 $\downarrow \qquad \qquad \downarrow \qquad \qquad \downarrow$
 $H_k(X^v,X^u) \stackrel{\bar{\delta}_{k-1}''}{\to} H_{k-1}(C^v,C^u) \to H_{k-1}(A^v,A^u) \oplus H_{k-1}(B^v,B^u) \to H_{k-1}(X^v,X^u)$
 $\downarrow \qquad \qquad \downarrow \qquad \qquad \downarrow$
 $H_k(X,X^u) \stackrel{\bar{\delta}_{k-1}}{\to} H_{k-1}(C,C^u) \to H_{k-1}(A,A^u) \oplus H_{k-1}(B,B^u) \to H_{k-1}(X,X^u)$

 $Rel_{\nu}^{u,v}(X) = Rel_{\nu}^{u,v}(A) + Rel_{\nu}^{u,v}(B) - Rel_{\nu}^{u,v}(C) + \operatorname{rk}\bar{\delta}_{\nu}' - \operatorname{rk}\bar{\delta}_{\nu-1}'' + \operatorname{rk}\bar{\delta}_{\nu-1}$

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Mayer-Vietoris formula for extended persistence (u < v

• For every $u < v \in \mathbb{R}$, and $k \in \mathbb{Z}$

$$\mathsf{Ext}_k^{u,\mathsf{v}}(X) = \mathsf{Ext}_k^{u,\mathsf{v}}(A) + \mathsf{Ext}_k^{u,\mathsf{v}}(B) - \mathsf{Ext}_k^{u,\mathsf{v}}(C) + \mathrm{rk}\hat{\delta}_k' - \mathrm{rk}\hat{\delta}_k'' + \mathrm{rk}\hat{\delta}_{k-1}$$

with $\hat{\delta}_k'$, $\hat{\delta}_k''$, and $\hat{\delta}_{k-1}$ as in the following diagram:

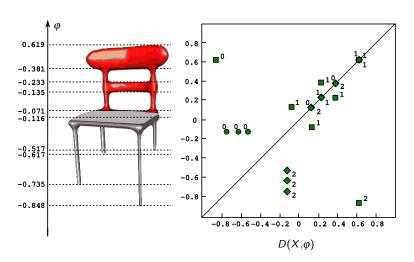
Mayer-Vietoris formula for extended persistence ($v \le N$

• For every $v \leq u \in \mathbb{R}$, and $k \in \mathbb{Z}$

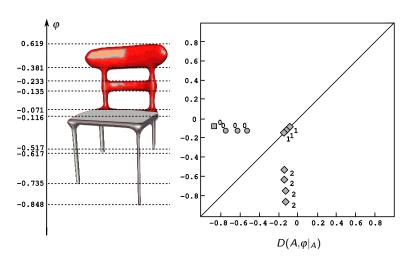
$$Ext_k^{u,v}(X) = Ext_k^{u,v}(A) + Ext_k^{u,v}(B) - Ext_k^{u,v}(C) + \operatorname{rk}\widetilde{\delta}'_k - \operatorname{rk}\widetilde{\delta}''_{k-1} + \operatorname{rk}\widetilde{\delta}_{k-1}$$

with $\widetilde{\delta}'_k$, $\widetilde{\delta}''_{k-1}$, and $\widetilde{\delta}_{k-1}$ as in the following diagram:

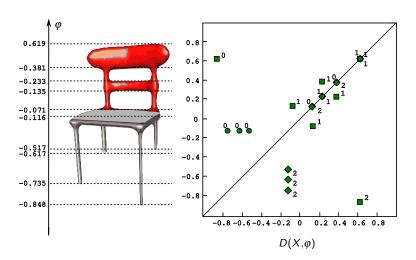




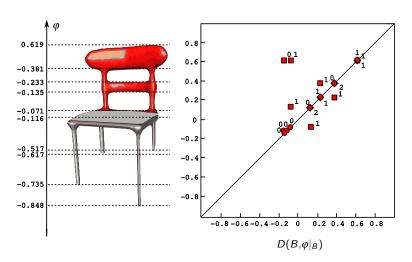














Experimental results





