# HIERARCHICAL CLUSTERING

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# Hierarchical clustering

Instead of partitioning the input set X, we compute a **hierarchy**.



Wlements that are "similar" should be close in the tree.

One very popular approach is agglomerative clustering.

As usual suppose we have a **distance**  $d : X \times X \mapsto \mathbb{R}^+_0$ .

**Algorithm 1:** AgglomerativeClustering(X, d)

```
for each x \in X do create a cluster C_x = \{x\};
while there are at least 2 clusters left do
find the two "closest" clusters C_i and C_j;
create a cluster C_i \cup C_j and set it as parent of C_i and C_j
end
```

return the resulting tree;

# Agglomerative clustering

The algorithm is not well-defined until we specify what "closest clusters" means. **Single-linkage** a.k.a. **nearest neighbor** 

 $d(C, C') = \min_{x \in C} \min_{y \in C'} d(x, y)$ 

#### Complete-linkage a.k.a. farthest neighbor

 $d(C, C') = \max_{x \in C} \max_{y \in C'} d(x, y)$ 

Ward

$$d(C,C') = \sum_{\bm{x} \in C \cup C'} \|\bm{x} - \bm{\mu}(C \cup C')\|_2^2 - \left(\sum_{\bm{x} \in C} \|\bm{x} - \bm{\mu}(C)\|_2^2 + \sum_{\bm{x} \in C'} \|\bm{x} - \bm{\mu}(C')\|_2^2\right)$$

# Example: Ward clustering of European languages

```
langs <- read.csv("langfreq.csv", row.names = 1)
D <- dist(langs)
hc <- hclust(D, method = "ward.D2")
plot(as.dendrogram(hc))
cutree(...) # try</pre>
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