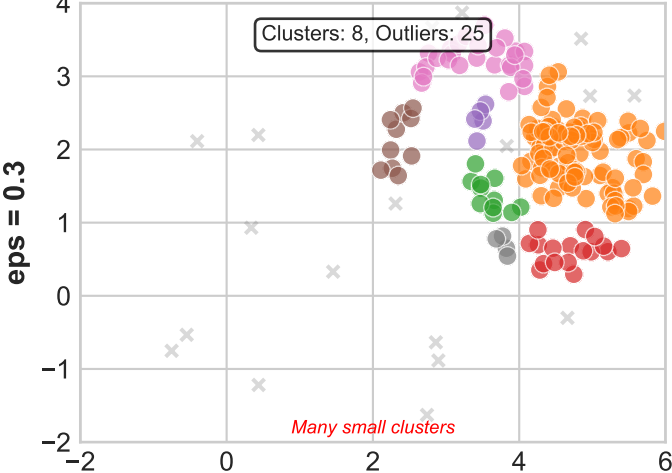
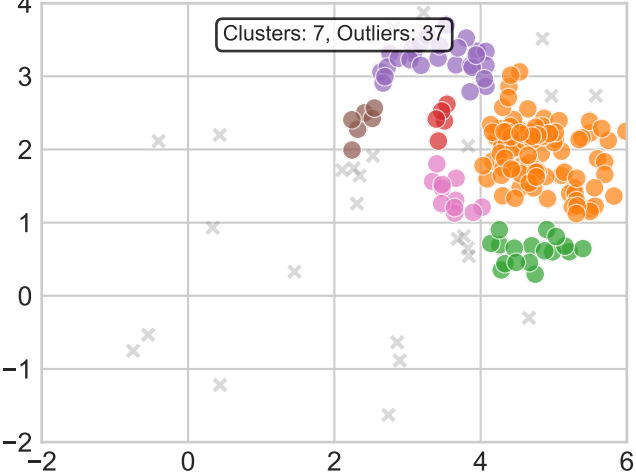


# DBSCAN Parameter Tuning: Impact on Innovation Clustering

eps=0.3, min\_samples=3  
min\_samples = 3



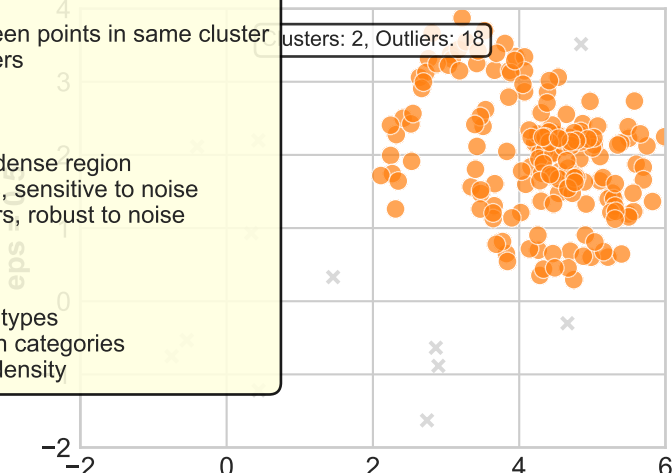
eps=0.3, min\_samples=5  
min\_samples = 5



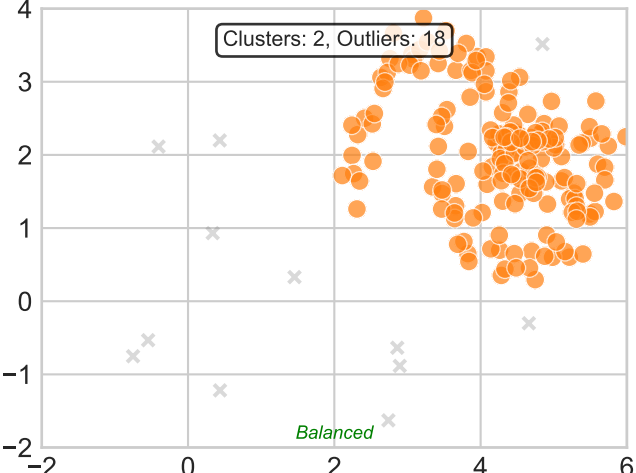
eps=0.3, min\_samples=10  
min\_samples = 10



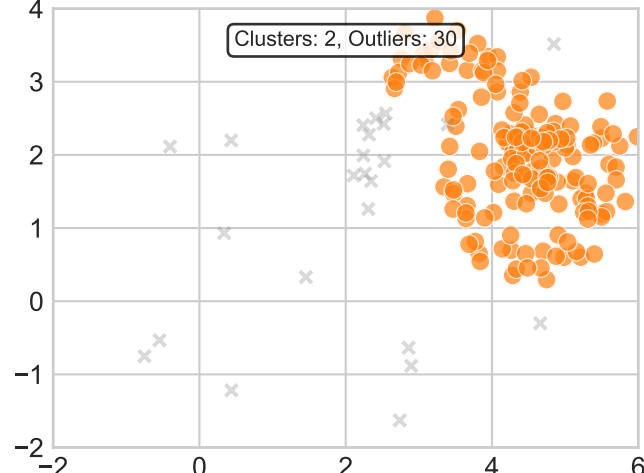
eps=0.5, min\_samples=3



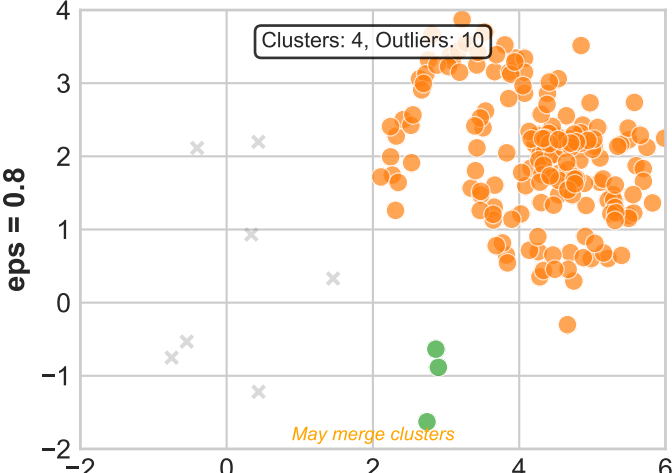
eps=0.5, min\_samples=5



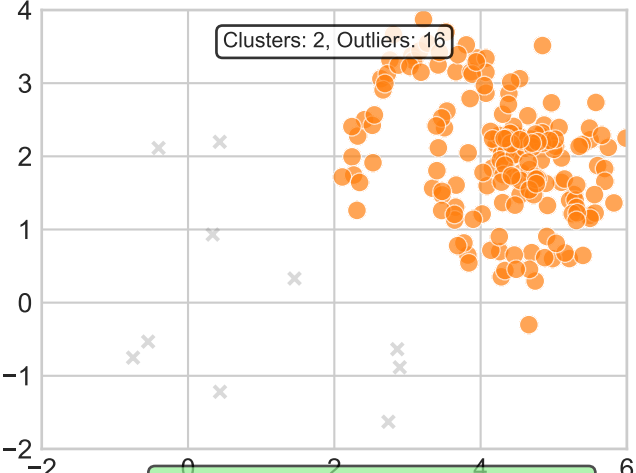
eps=0.5, min\_samples=10



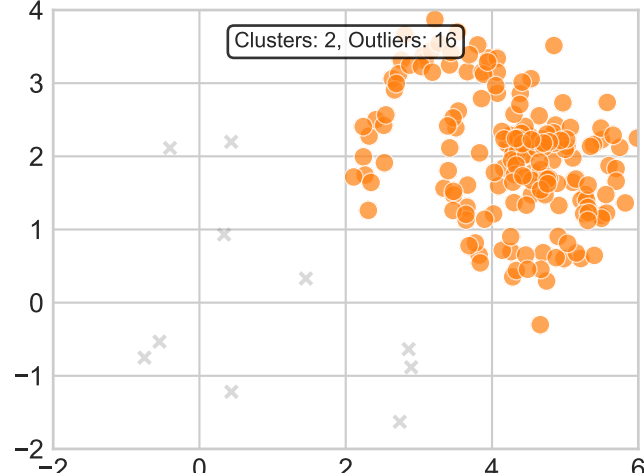
eps=0.8, min\_samples=3



eps=0.8, min\_samples=5



eps=0.8, min\_samples=10



Innovation Clustering Context:  
\* Dense areas = Mainstream innovations  
\* Sparse areas = Radical innovations  
\* Outliers = Breakthrough ideas

Parameter Guidelines:

eps (epsilon): Maximum distance between points in same cluster

- \* Small eps → Many small, tight clusters
- \* Large eps → Fewer, larger clusters
- \* Too large → All points in one cluster

min\_samples: Minimum points to form dense region

- \* Small min\_samples → More clusters, sensitive to noise
- \* Large min\_samples → Fewer clusters, robust to noise
- \* Too large → Many outliers

For Innovation Data:

- \* Use small eps for distinct innovation types
- \* Use larger eps for broader innovation categories
- \* Adjust min\_samples based on data density

Tuning Strategy:

1. Start with k-distance plot
2. Look for 'elbow' in plot
3. Set eps at elbow point
4. min\_samples = 2\*dimensions
5. Validate with domain knowledge