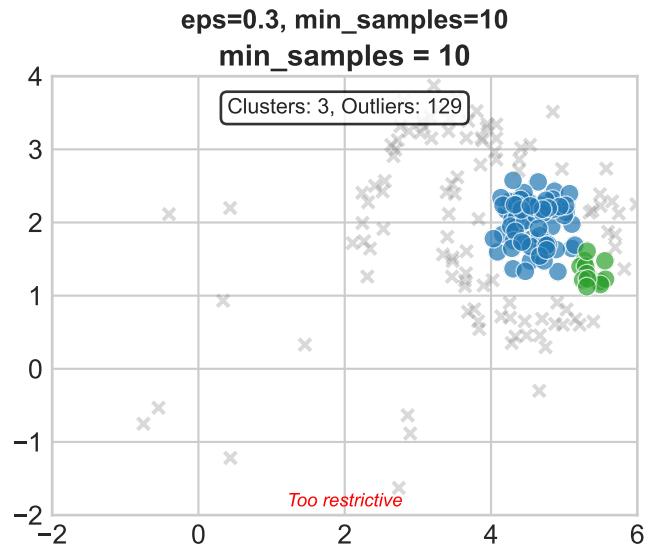
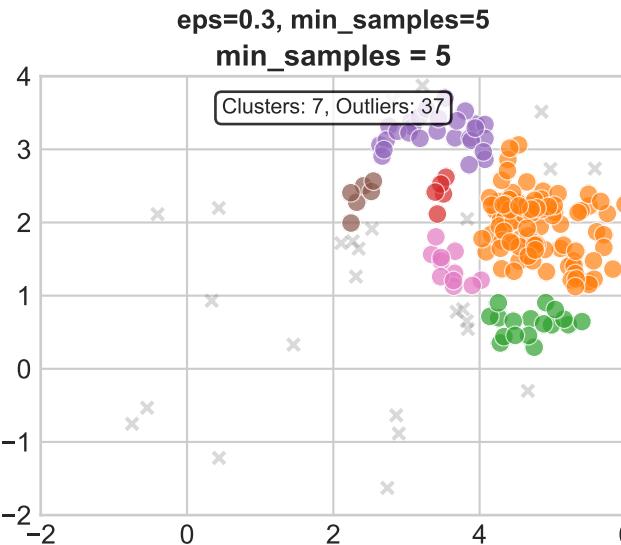
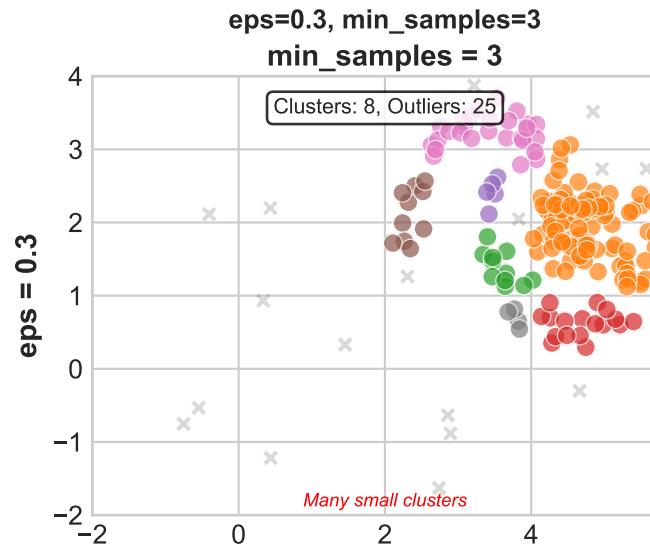


# DBSCAN Parameter Tuning: Impact on Innovation Clustering



**Parameter Guidelines:**

**eps**: Maximum distance between points in same cluster

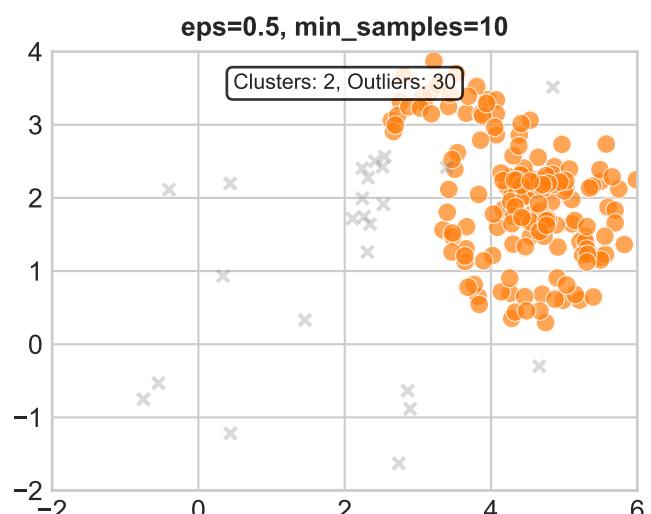
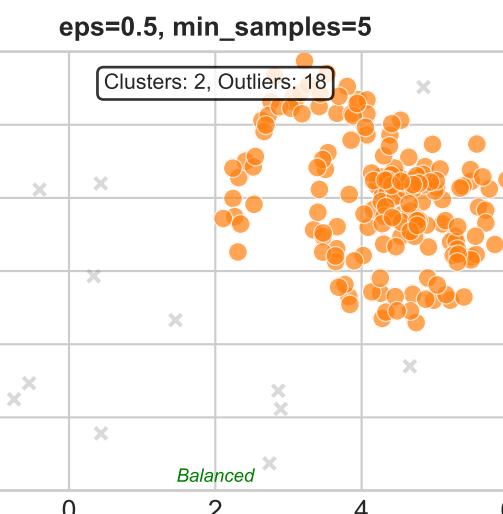
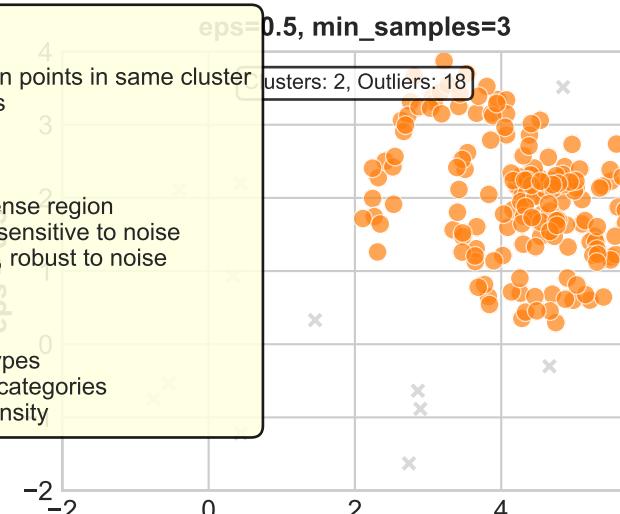
- \* Small eps → Many small, tight clusters
- \* Large eps → Fewer, larger clusters
- \* Too large → All points in one cluster

**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

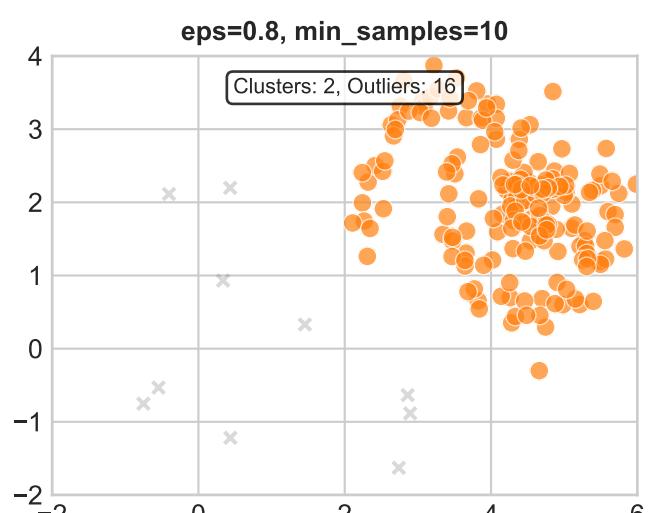
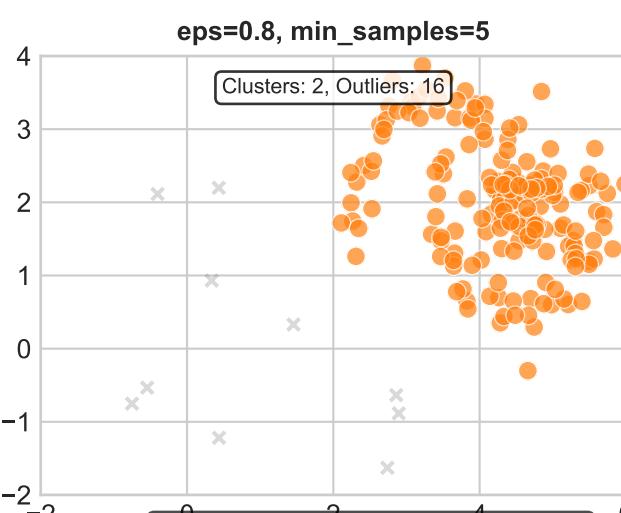
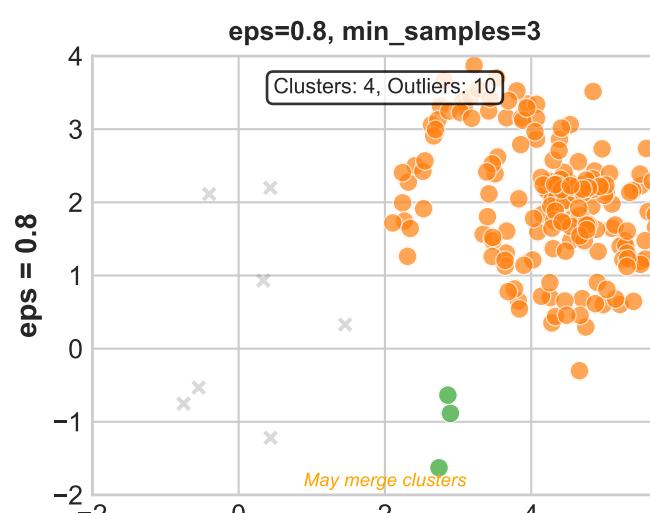
**For Innovation Data:**

- \* 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



**Innovation Clustering Context:**

- \* Dense areas = Mainstream innovations
- \* Sparse areas = Radical innovations
- \* Outliers = Breakthrough ideas