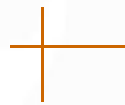
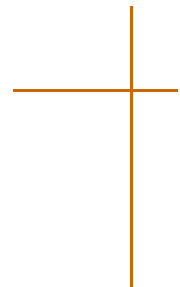


Data Analytics for Transmission Expansion Planning Estadística II

Andrés Ramos
January 2018



Introduction



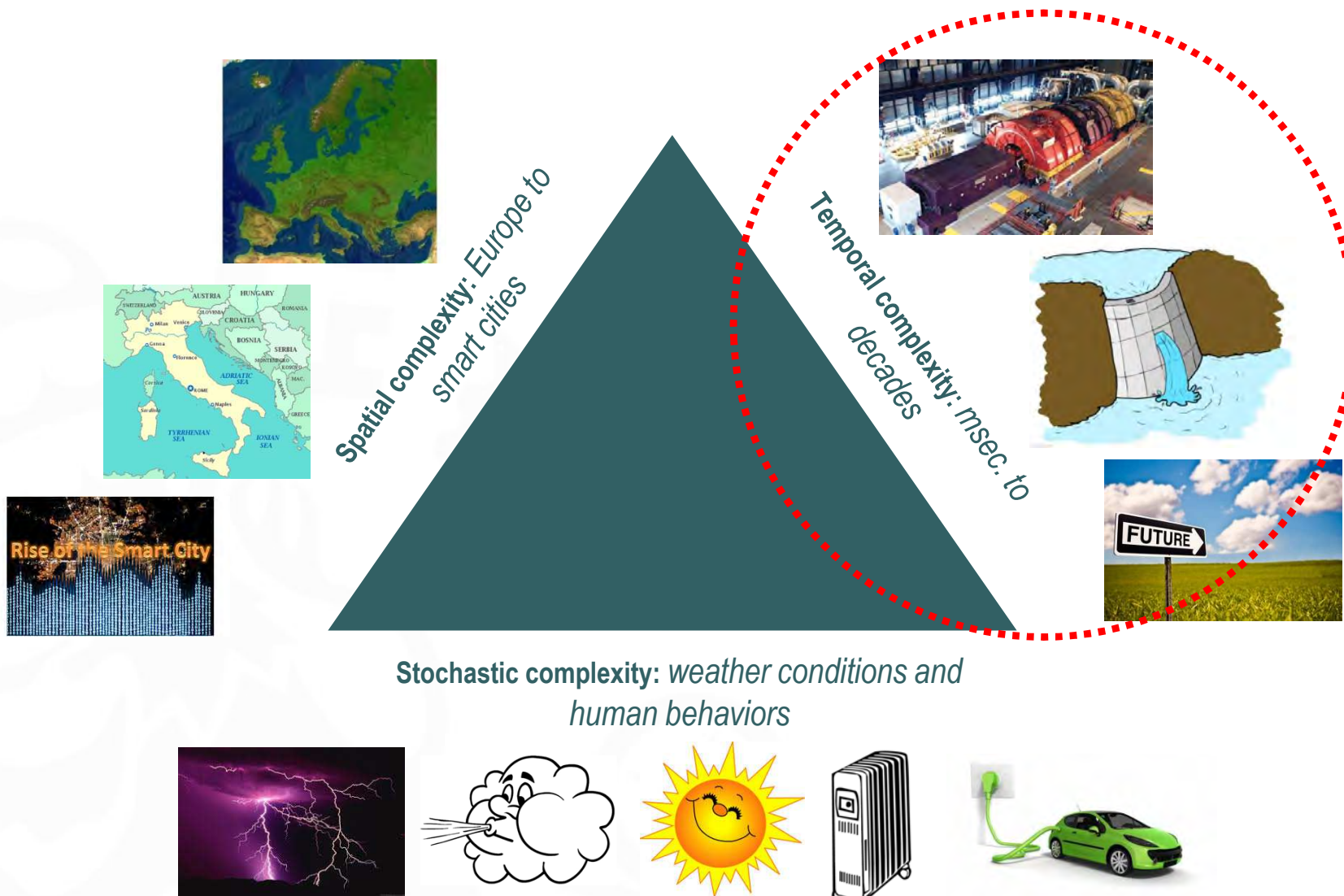
Transmission Expansion Planning

Definition

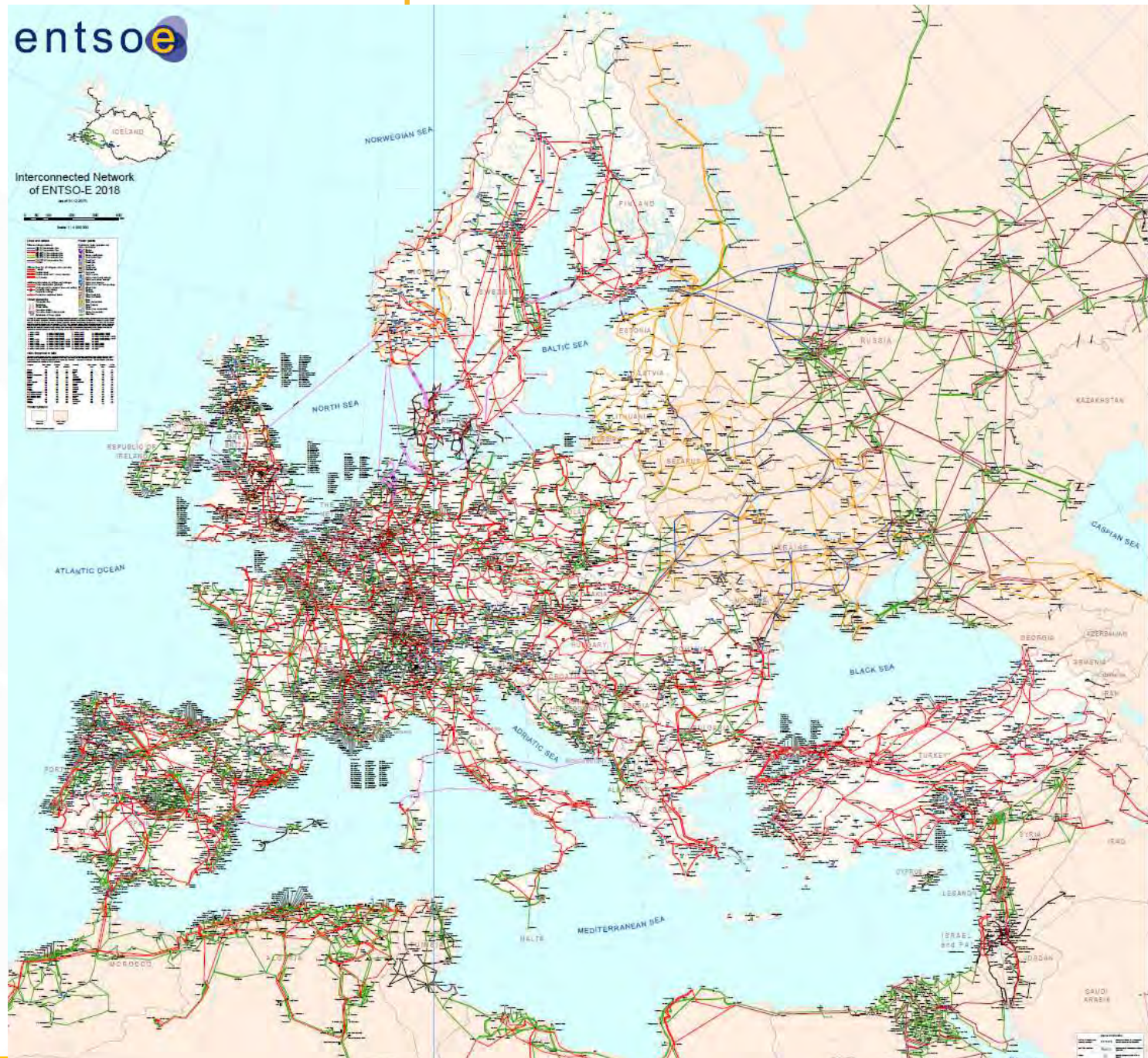
- Determine **which** lines and transformers and **when** to build **optimizing total investment and operation costs**



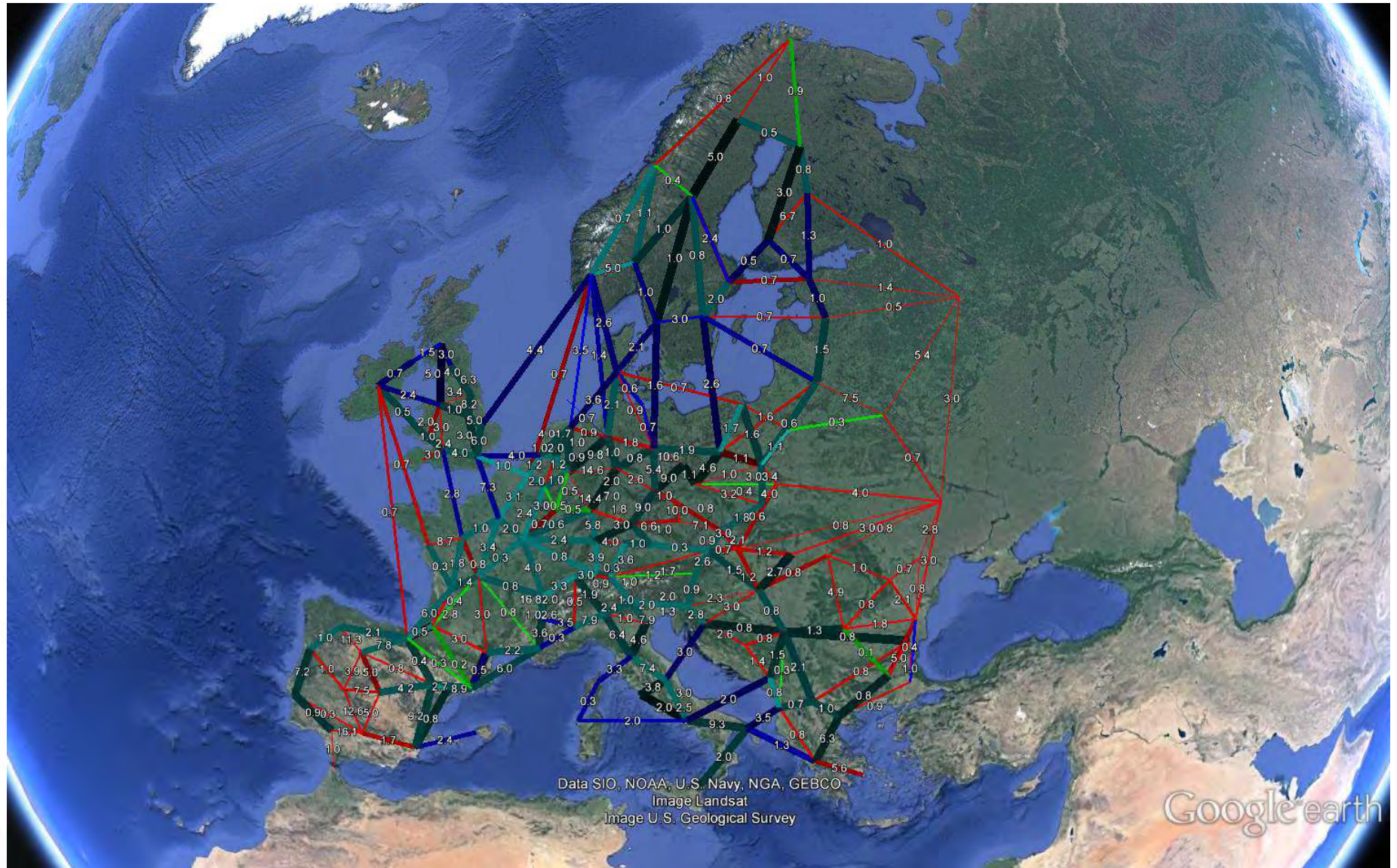
Challenges for TEP



ENTSO-E Grid Map 2018

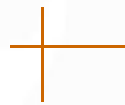
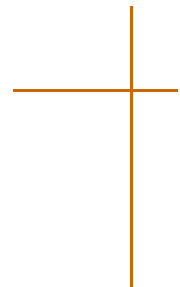


TEP for Europe





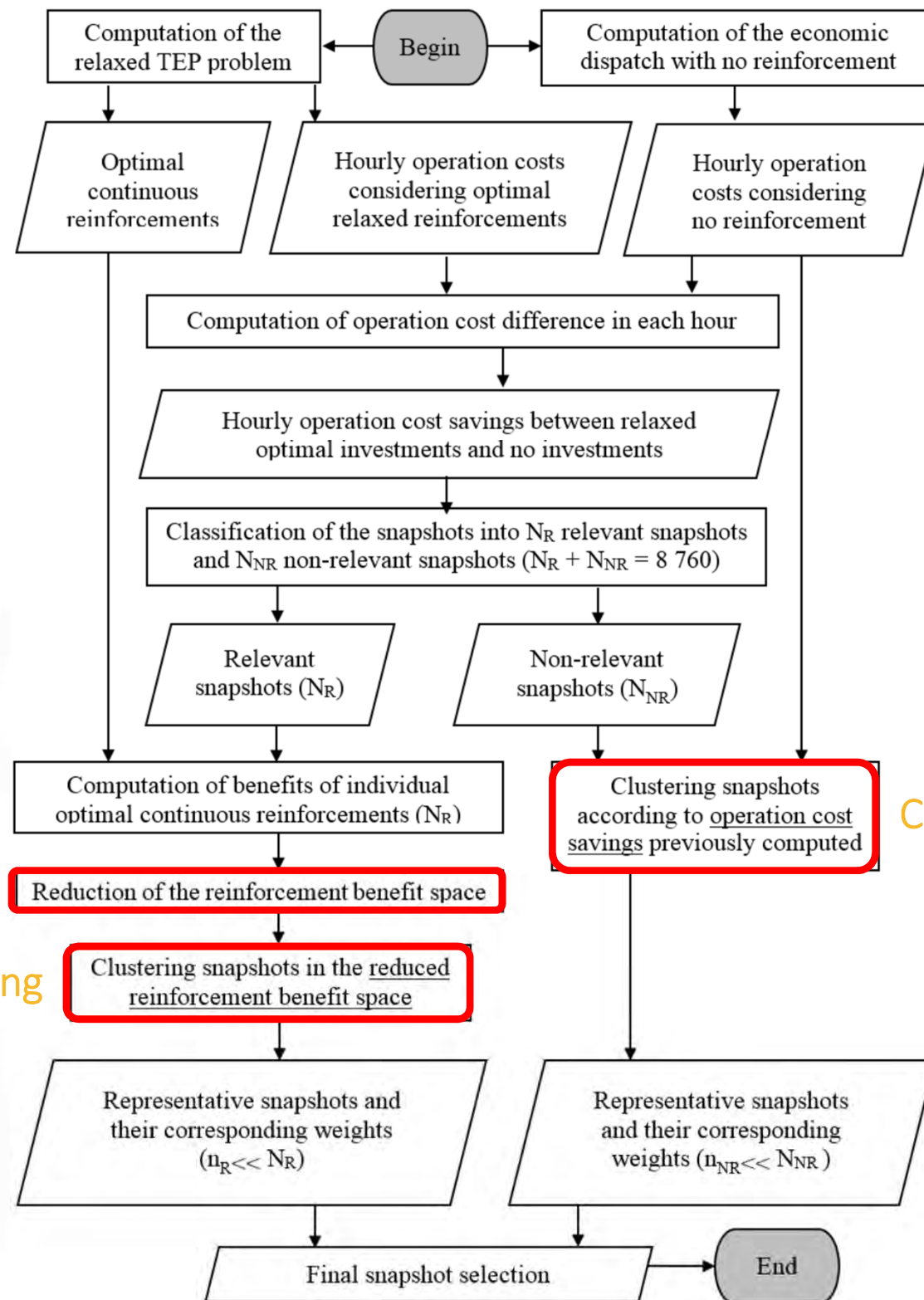
Data Analytics for TEP



Reference

- Q. Ploussard, L. Olmos and A. Ramos [*An operational state aggregation technique for transmission expansion planning based on line benefits*](#) IEEE Transactions on Power Systems 32 (4): 2744-2755 Oct 2017
[10.1109/TPWRS.2016.2614368](#)

Overview



PCA

Clustering

Clustering

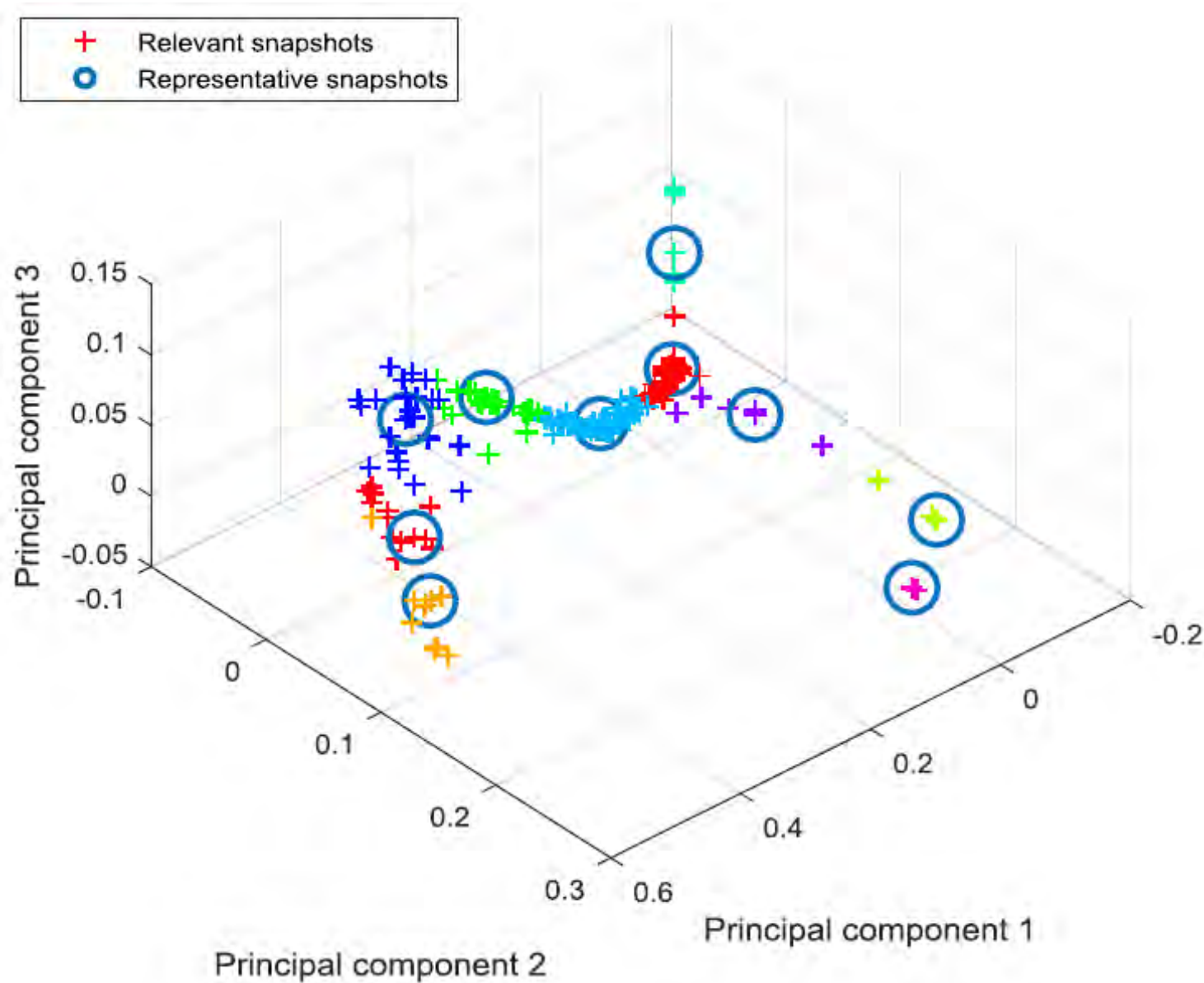
Principal Component Analysis (PCA)

- In order to **reduce the dimension of the benefit space**, Principal Component Analysis (PCA) is applied to original reinforcement benefits.
- Transforms a dataset in a new one that provides the same information as the original one but in a more compact way. The transformation of the original dataset is defined in such a way that the first principal component of the transformed dataset has the largest possible variance (that is, accounts for as much of the variability in the dataset as possible), and each succeeding component has, in turn, the highest variance possible under the constraint that it is orthogonal to the preceding components. The major part of the variability across operation situations existing in the initial reinforcement benefit space is captured using the first Principal Components.
- Thanks to this process, the dimension of the **reinforcement benefit space can be drastically reduced** without incurring an important loss of relevant information about the operation situations to be clustered.

Clustering

- The **K-means** algorithm is applied to **cluster operation situations within the reduced reinforcement benefit space (comprising the principal components of the original reinforcement benefit space)**. The **Euclidean distance** is used to compare operation situations. There are different possible criteria to measure the goodness of a clustering of samples (operation situations) for a given number of clusters K . The one used aims to minimize the sum of distances of samples to their associated centroid (center of their cluster), i.e., the **intra-cluster distance**.
- Most clustering algorithms start from an initial set of K samples, defined as the **initial set of K centroids, and iteratively try to increase the goodness** of the clustering of operation situations by modifying the set of centroids. Improving the initial choice of centroids leads to computing a good partition of snapshots in fewer iterations and avoiding trivial partitions (those where some clusters contain only one snapshot). The **K-means++ algorithm** addresses this concern by **increasing the spread of the initial set of centroids**. A first centroid is chosen randomly, considering a uniform distribution for the probability of choosing any snapshot as the first centroid. Then, each subsequent centroid is chosen among the remaining snapshots when each of these snapshots is assigned a probability of being chosen proportional to the square distance in the reduced benefit space from this snapshot to the closest snapshot already chosen as a centroid. This process ends when the initial set of K centroids are chosen.

Clustering relevant snapshots considering 10 clusters



*Thank you for your
attention*

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