Visualization methods combined to clustering approach enabled the identification of clinically relevant patterns of prostate cancer management. deaux Pharm

ABSTRACT

BACKGROUND

The heterogeneity of prostate cancer (PC) patient journeys for a same stage of disease makes their evaluation complex using descriptive statistics. Unsupervised machine learning has the potential to reveal patterns within heterogeneous data. However, impact of such methods in real-world studies is not yet clear.

OBJECTIVE

To illustrate how clustering and visualization of healthcare pathways can enhance the characterization of patients with PC, at all disease

METHODS

Data source & Tools:

- · French nationwide healthcare database (Système National des Données de Santé, SNDS)
- R packages used: clustofVar, TraMineR WeightedCluster

Study population & Data:

- Participants: Patients with prevalent prostate cancer in 2014 were identified among men alive, aged ≥40, and covered by the general health insurance scheme in 2014 based on discharge diagnosis and specific encounters
- Observation period: Data were extracted with a 5-year history period, and with up to 7 semesters of follow-up from the PC diagnosis
- A validated algorithm* was applied to detect castration resistance, metastasis management and so metastatic and castration-resistant status
- Four exclusive cohorts of incident patients were constituted, prioritizing the most advance disease stage:
 - ✓ hormone-sensitive (HSPC),
 - metastatic hormone-sensitive (mHSPC),
 - castration-resistant (CRPC)
 - metastatic castration-resistant (mCRPC).

Study stages and Clustering method: See beside

RESULTS

Clustering allowed to distinguish trends in healthcare pathways such as patients:

- Undergoing local treatment with or without androgen deprivation therapy in HSPC,
- With rapid and slow disease progression in
- With curative, or palliative intent treatment in mHSPC and mCRPC.

CONCLUSION

Visualization methods combined to clustering approach enabled the identification of clinically relevant patterns of prostate cancer management. Characterization of these care pathways is an essential element for the comprehension and the robust assessment of healthcare technologies

*Thurin NH, Rouyer M, Gross-Goupil M, et al. Epidemiology of metas castration-resistant prostate cancer. A first estimate of incidence and prevalence using the French nationwide healthcare database. Cancer Epidemiol. 2020;

The authors received no financial support for the research. However, the example used was drawn from the TherapeutiC strategy in Metastatic castration-Resistant pRostate cAncer: target population and changes between 2012 and 2014 (CAMERRA) study, which was funded by Jan Cliag, France and carried out by the Bordeaux PharmacoEpi platform the supervision of a Scientific Committee.

Applying sequence clustering methods to characterize healthcare pathways of patients at different prostate cancer stages in the French nationwide healthcare database (SNDS)

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Study stages & Clustering method

51 specific healthcare encounters constitutive of PC management were synthetized into 4 macro-variables: surveillance, local treatment, androgenic deprivation, and advanced treatment based on the whole patient data.

For each patient, 1 state was assigned per semester among 16 possible ones according to the values of macro-variables, succession of these states over time formed the health



In each cohort (HSPC, mHSPC, CRPC, mCRPC) optimal matching (TRATE substitution method) was applied to calculate distances between pathways, and Partitioning Around Medoids (PAM) algorithm was used to generate consistent groups of similar pathways.

State Sequence Analysis (SSA) method was used to visualize clustered healthcare pathways. Optimal number of clusters was selected from internal validation indexes (Hubert coefficient, average silhouette width...).

Healthcare states

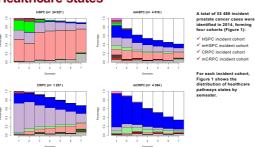
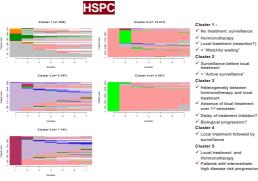




Figure 1. Distribution of healthcare states by semester among the four incident cohorts

Patients pathway



re 2. Cluster index plots of the incident HSPC cohort (n = 24 927)

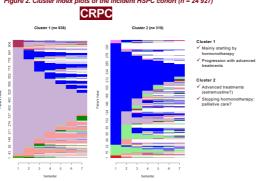
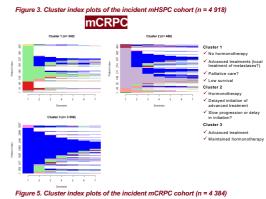


Figure 4. Cluster index plots of the incident CRPC cohort (n = 1 257)







ISPOR Europe 2022 Vienna, Austria, Nove









