

## Ensemble Clustering to Generate Phenotypes of Kidney Transplant Donors and Recipients

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## Background

- Kidney transplantation is a treatment option for end-stage kidney disease
- Donor and recipient characteristics impact outcomes for kidney transplant
- Understanding the relationships between donor and recipient characteristics
- Understanding the relationships between donor and recipient characteristics and how they influence post transplant outcomes may have future implications for kidney allocation



## Objectives and Approach

- To identify phenotypes of kidney transplant recipients using their clinical characteristics (inclusive of both donor and recipient features)
- To determine how these phenotypes influence graft survival.
- Investigate a dataset of kidney transplant donor and recipient characteristics
- Generate phenotypes by applying unsupervised clustering to multi-type data



## Data

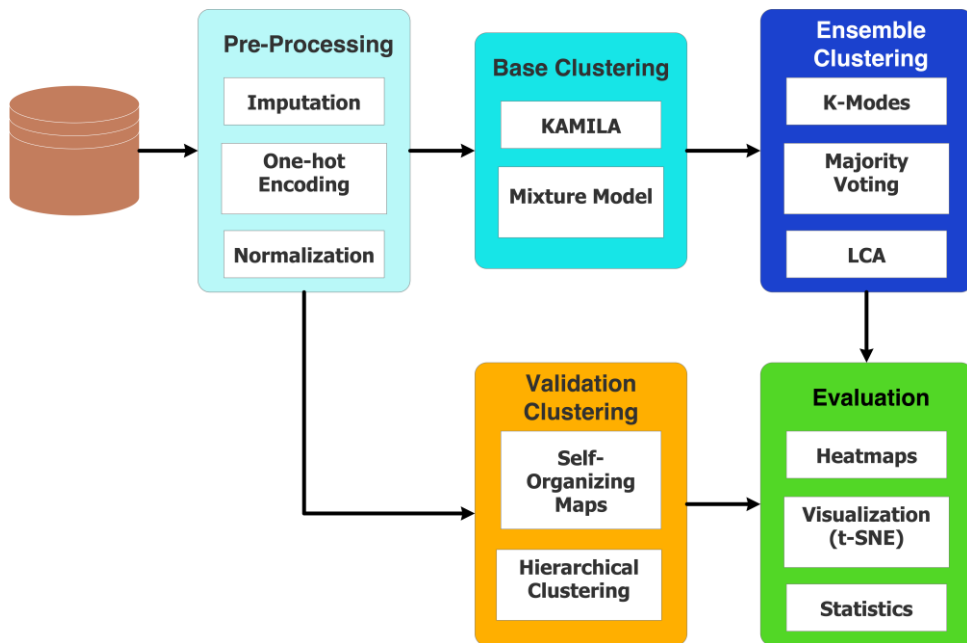
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- Scientific Registry of Transplant Recipients (SRTR)—2009 to 2011
- 25,824 recipients of a deceased donor kidney transplant
- 28 numerical and categorical variables
- Recipient outcome (experienced graft failure, died or survived) not used



## Methods

- Multi-type data clustering
  - Model based clustering: Mixture models and Kay-means for Mixed Large data (KAMILA)
  - Neural network clustering: Self Organizing Maps
- Ensemble clustering
- Cluster evaluation metrics
  - Silhouette score, Dunn index, Calinski-Harabasz score





## Results – Base vs Ensemble Clustering

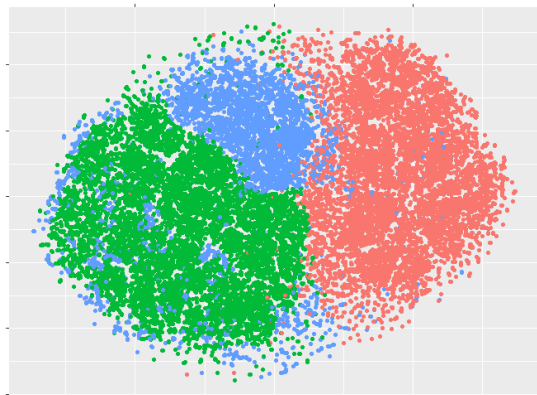
Method	Silhouette Score (↑)	Dunn Index (↑)	CH Index (↑)
Base Clustering			
KAMILA	0.108	0.0536	2760
Mixture model	0.091	0.0099	2514
Ensemble Clustering			
k-modes	0.108	0.0200	2704
Majority Voting	0.076	0.0282	2065
Latent Class Analysis (LCA)	0.113	0.0538	2843
Validation Clustering			
Self-Organizing Map (SOM)	0.100	0.0512	2595



## Clustering Results

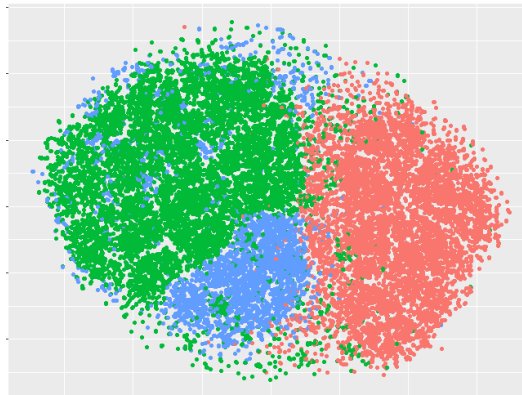
### • Latent Class Analysis

- Cluster 1: 11472
- Cluster 2: 9583
- Cluster 3: 4769

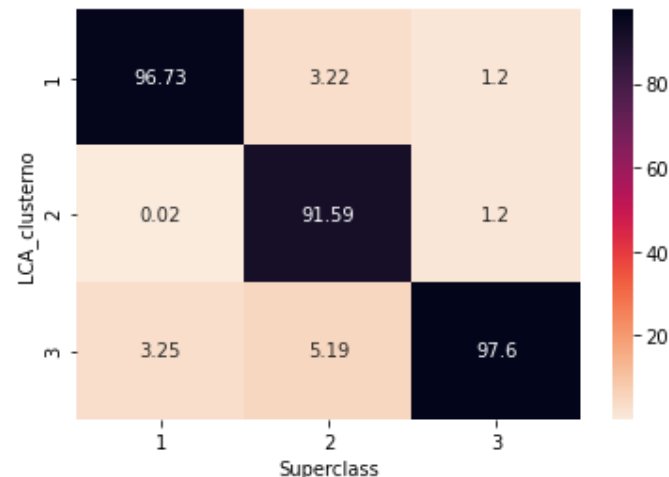


### • Self-Organizing Map

- Cluster 1: 12472
- Cluster 2: 9443
- Cluster 3: 3909



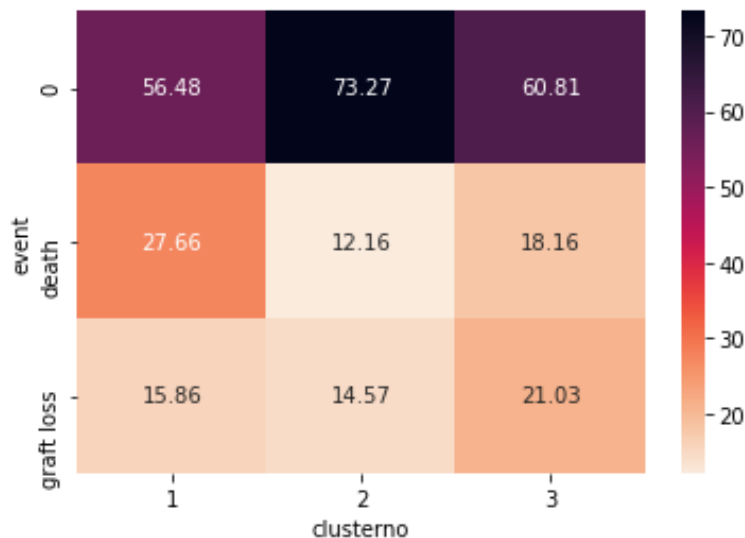
SOM



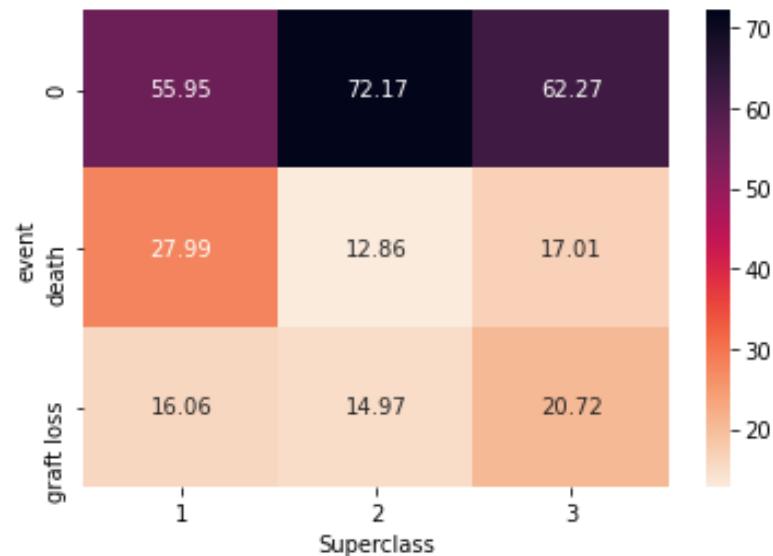


## Cluster Validation

LCA



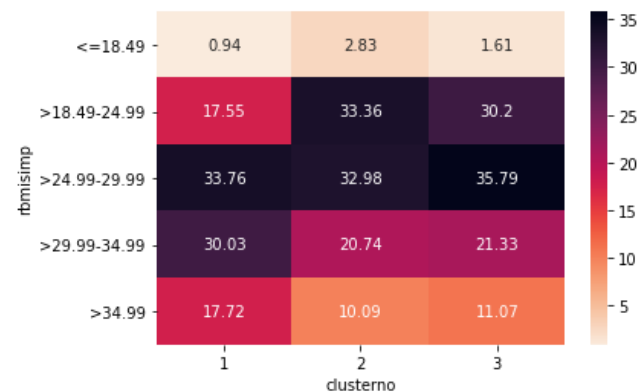
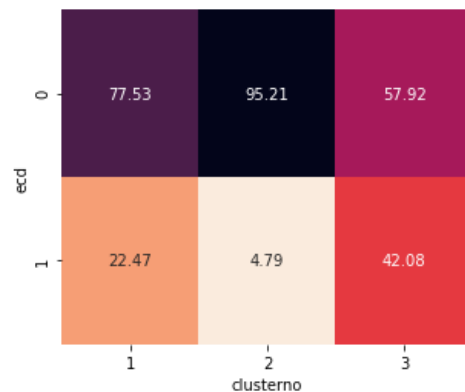
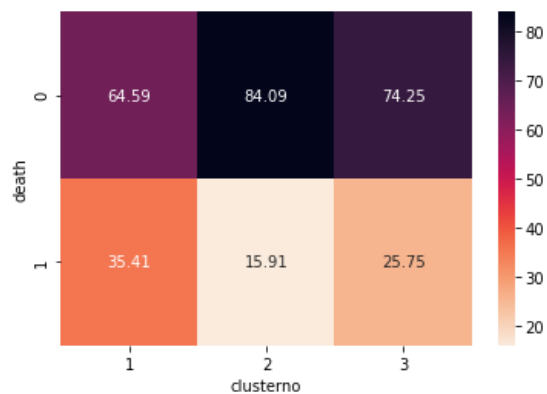
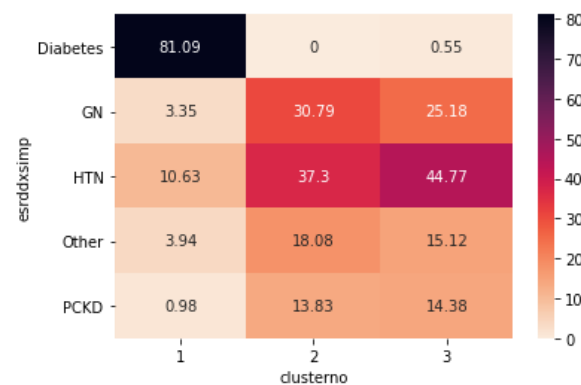
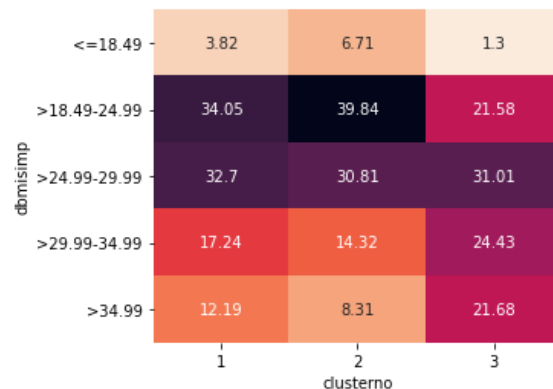
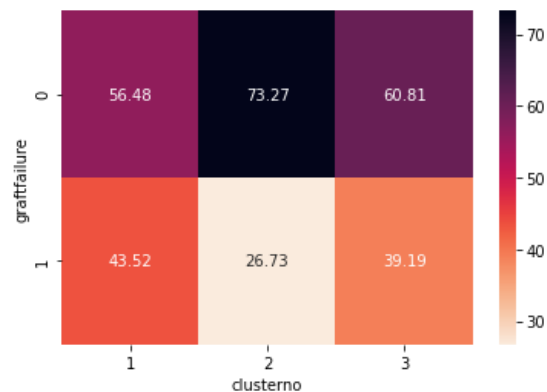
SOM





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Phenotype Feature	Cluster A (11,472)	Cluster B (9,583)	Cluster C (4,769)
<b>Donor &amp; Recipient Age</b>	Youngest Donor & Recipient	Oldest Recipient	Oldest D & Older R than cluster A
<b>Survival Time</b>	Longest	Similar	
<b>Years on dialysis pre-transplant</b>	Longer than cluster B	Shortest	Longest
<b>Recipient Event</b>	Highest censored	<b>Highest Death</b>	<b>Highest Graft Loss</b>
<b>Recipient Death</b>	Lowest	<b>Highest</b>	Higher than cluster A
<b>ESRD Diagnosis</b>	GN and HTN	Diabetes	GN and HTN
<b>Recipient Diabetes</b>	None	<b>Almost entirely Yes</b>	Mostly No
<b>Recipient Sex</b>	<b>Highest proportion of female sex</b>	Lowest	Higher than cluster B
<b>Recipient BMI</b>	Higher presence between 18.49 and 29.99 than cluster B	<b>Significantly higher &gt;29.99 and lower &gt;18.49 – 24.99</b>	Similar to cluster A
<b>Donor Hypertension</b>	Almost None	Majority No with significant Yes	<b>Almost entirely Yes</b>
<b>Expanded Criteria Donor</b>	Almost None	Majority No with significant Yes	<b>Highest Yes (~40%)</b>
<b>Donor Race</b>	Similar		<b>Highest proportion of black donors</b>
<b>Donor Diabetes</b>	Almost None	Slightly higher than cluster A	<b>Highest Yes</b>
<b>Donor BMI</b>	Highest >18.49 – 24.99	Higher presence >29.99 than cluster A	<b>Significantly higher &gt;29.99 and lower &gt;18.49 – 24.99</b>



## Concluding Remarks

- Phenotypes of kidney donors and recipients
  - Characterize anticipated outcomes for each cluster assignment
- Outcomes not used for cluster derivation
  - Clusters identify high risk phenotypes at differential risk of graft failure
- Multi-type data ensemble clustering approach
- Observed concordance between the clustering method provides a new strategy to validate cluster membership
- Generalizable to other organ transplant studies



## THANK YOU