Quantitative Viral Dynamics Across Scales, 2022

Principal Component Analysis for trajectories

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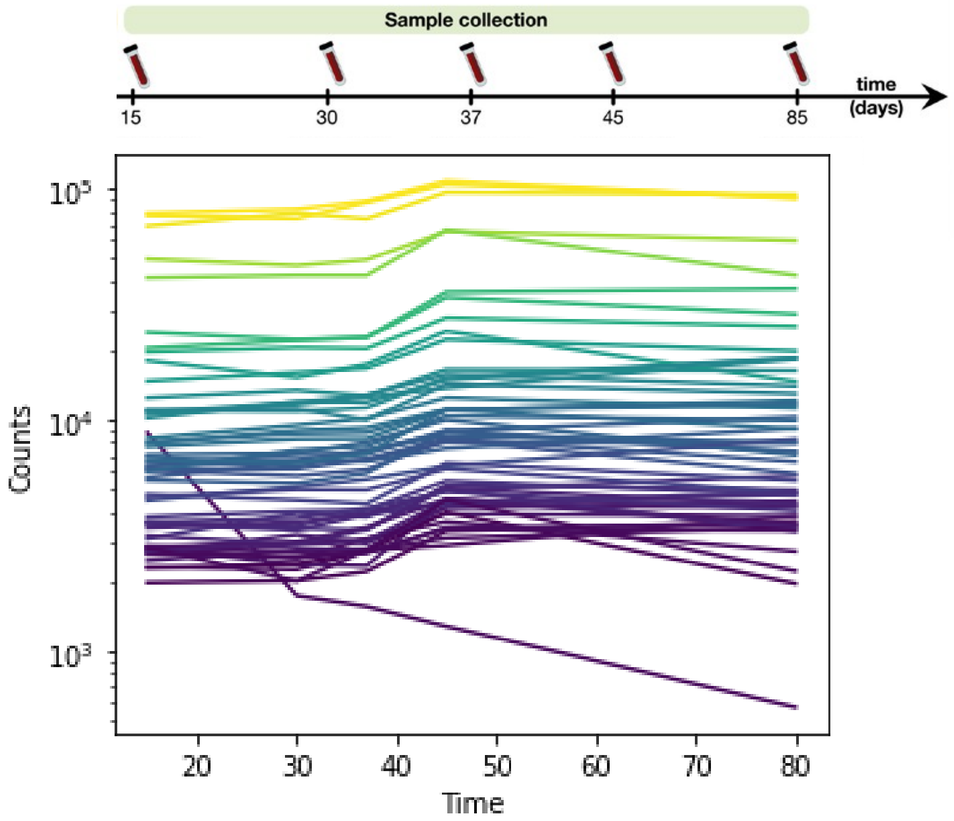
Why PCA for trajectories



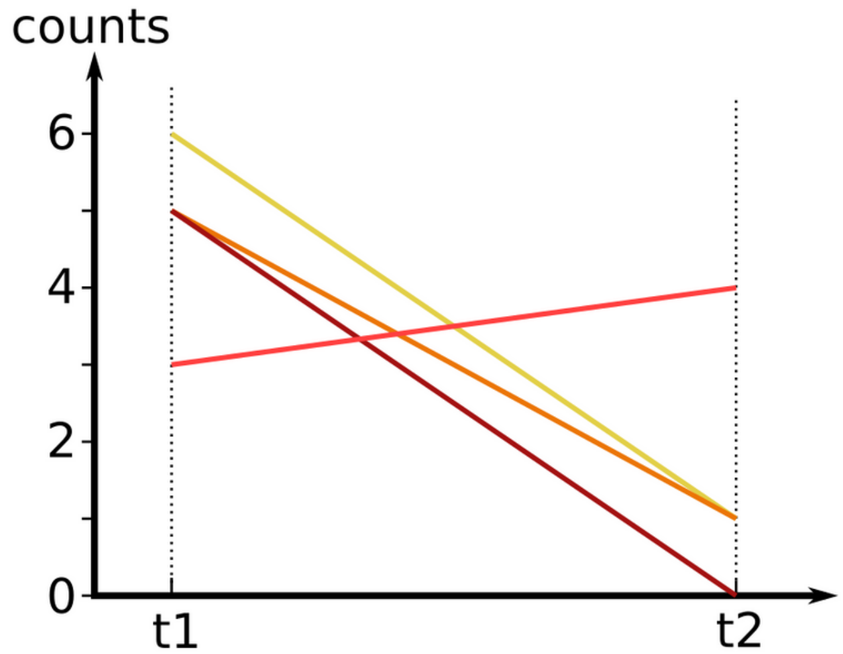
Our trajectory is a **5d** object (we have 5 time points)

We want to use PCA to do a **dimensional reduction** and represent the trajectory in a **2d** space

In this space we can more easily understand **global trends**

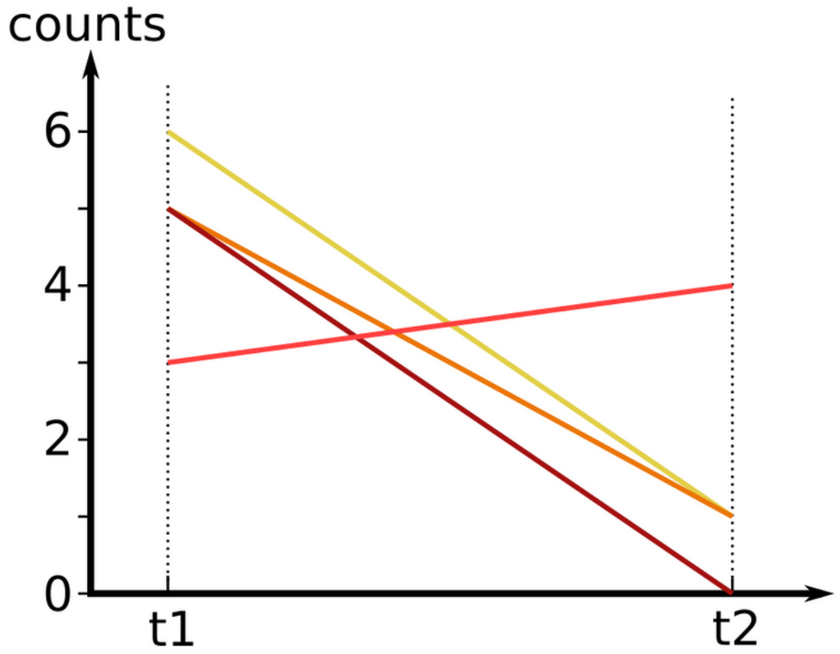
A toy example to understand PCA

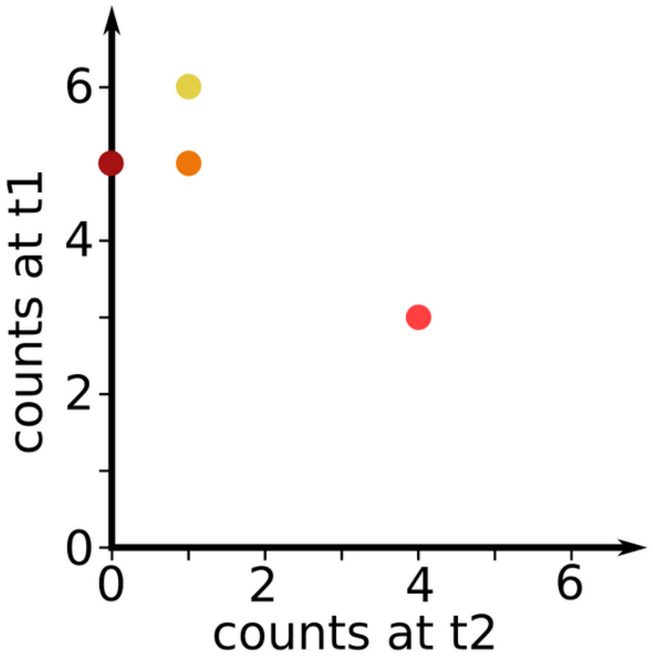
	Counts	
	t1	t2
CDR3 1	3	4
CDR3 2	5	0
CDR3 3	5	1
CDR3 4	6	1



A toy example to understand PCA

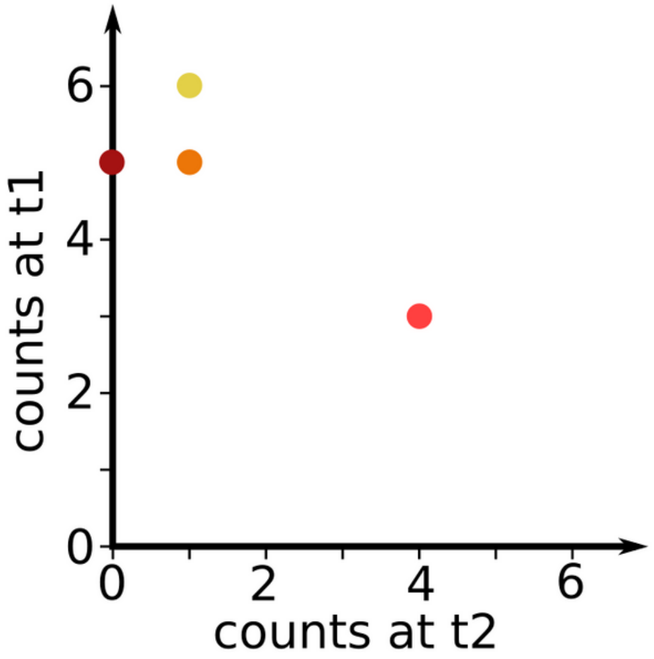
1 - Representation on the Cartesian plane

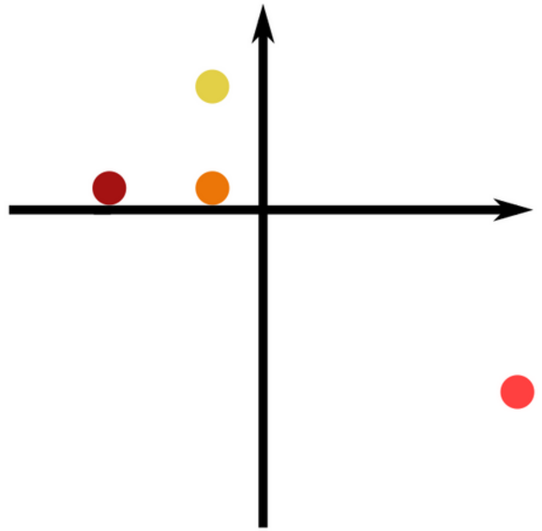




A toy example to understand PCA

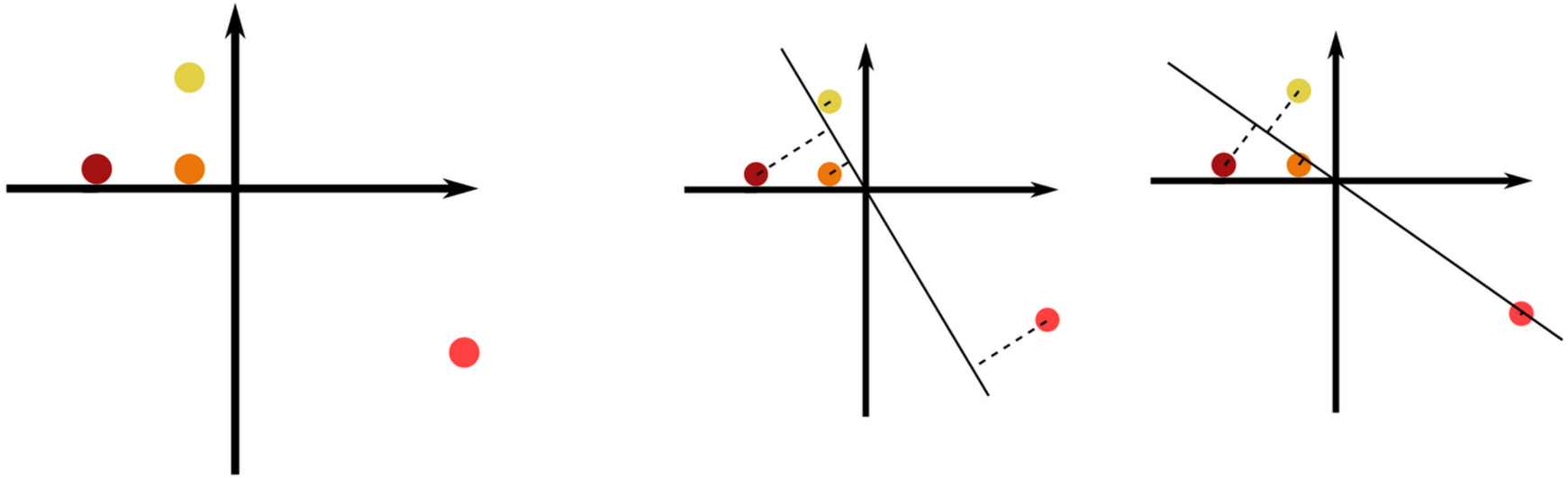
2 - Shift the origin to the center of mass





A toy example to understand PCA

3 - Find the best linear fit passing from the origin



i.e. minimization of the summation of the squared distances of the points from the line

A toy example to understand PCA

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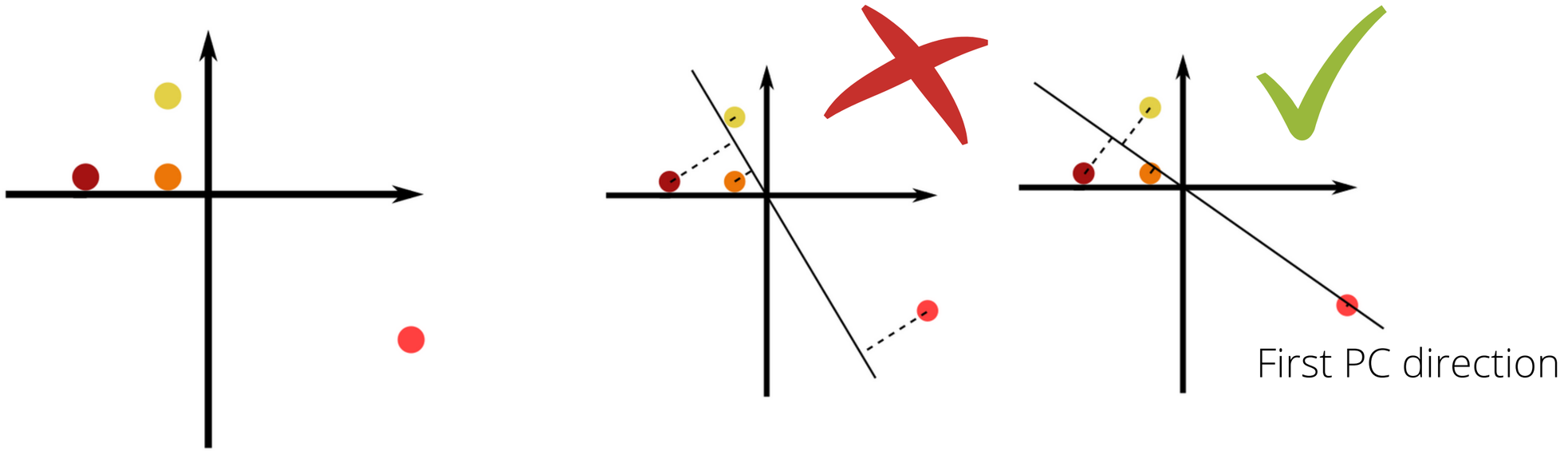
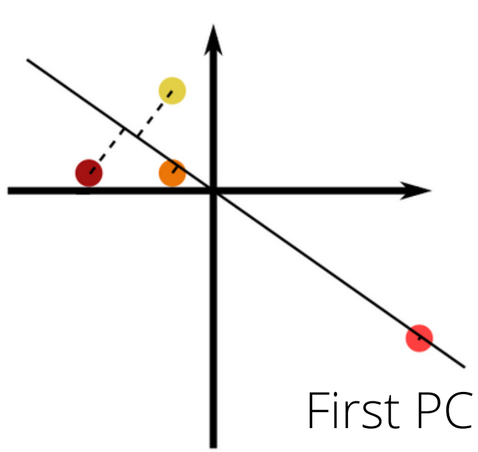


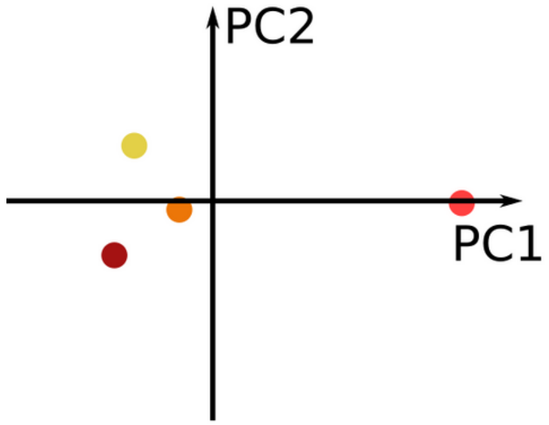
i.e. minimization of the summation of the squared distances of the points from the line

A toy example to understand PCA

4 - Rotation towards the first Principal Component

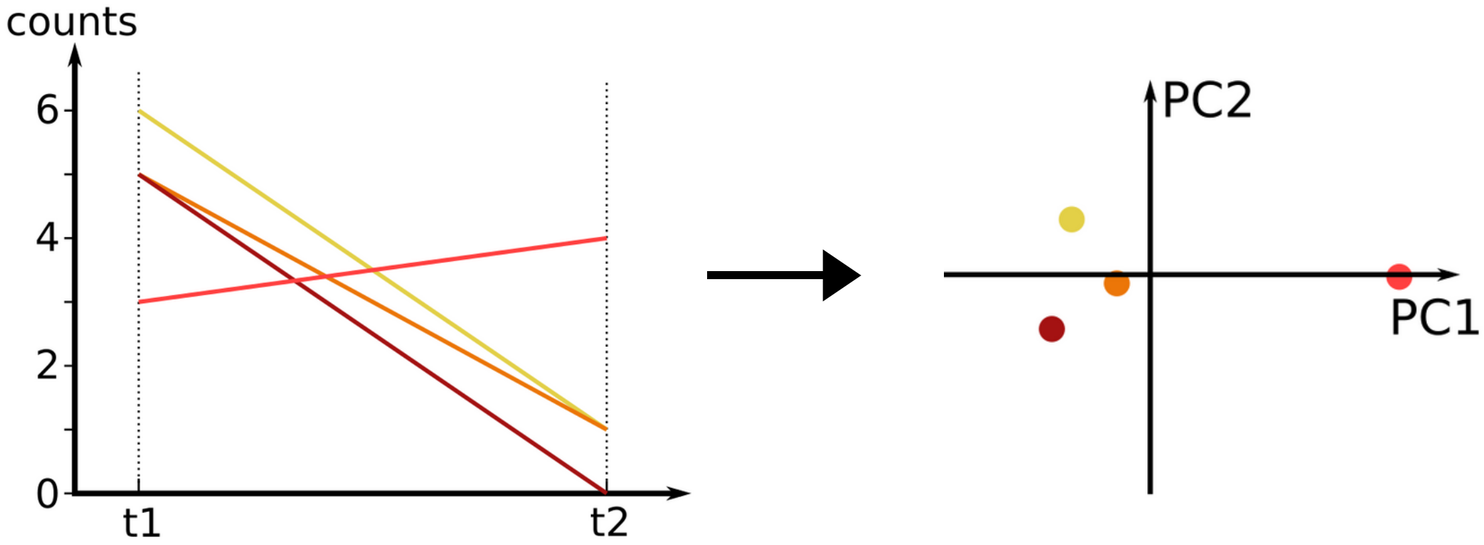


First PC direction



In 2d, the PC2 is the perpendicular line to PC1

A toy example to understand PCA

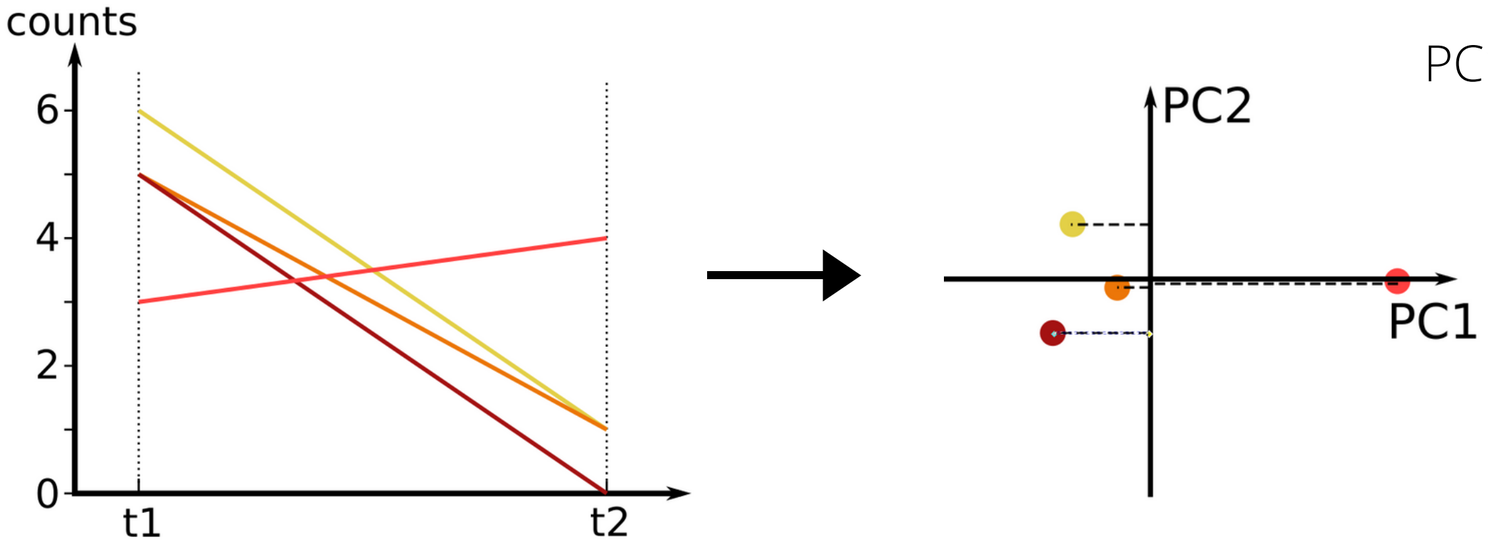


What we gain from this?

I'm ordering the axis from the **most informative** to the less

(true also for higher dimension)

A toy example to understand PCA



PC1 variation: 4.25

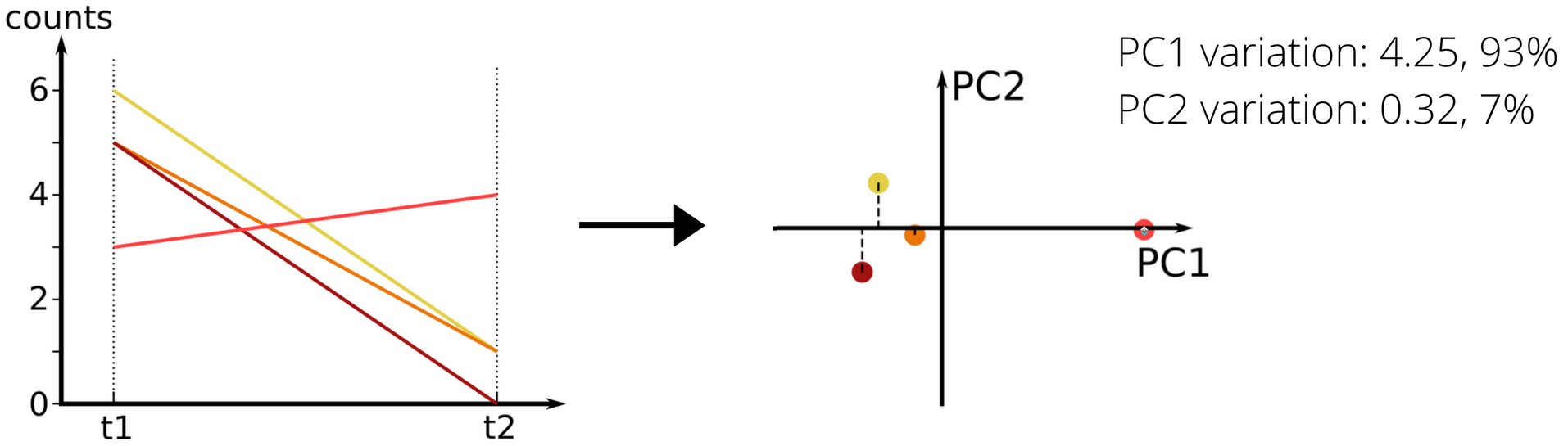
What we gain from this?

I'm ordering the axis from the **most informative** to the less

(true also for higher dimension)

Most informative: maximize data variation

A toy example to understand PCA



What we gain from this?

I'm ordering the axis from the **most informative** to the less

(true also for higher dimension)

Most informative: maximize data variation

A toy example to understand PCA

Extreme case: data are effectively 1d

	Counts	
	t1	**t2**
CDR3 1	2	4
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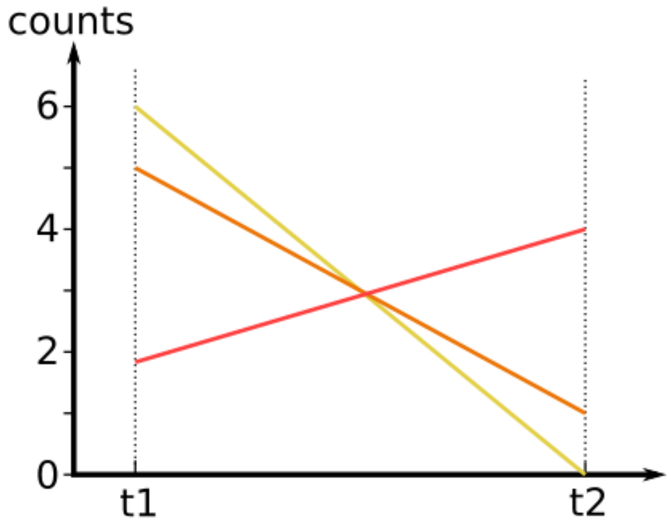
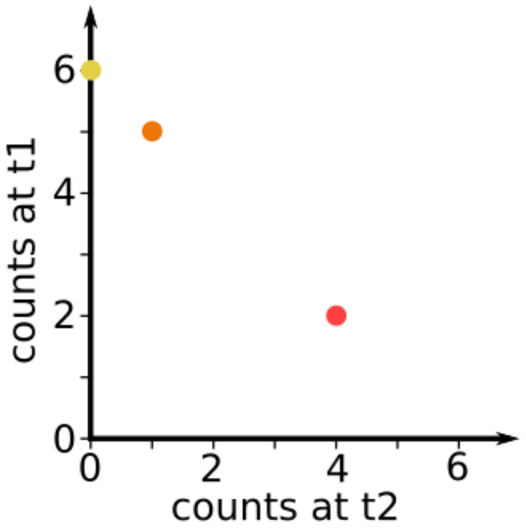
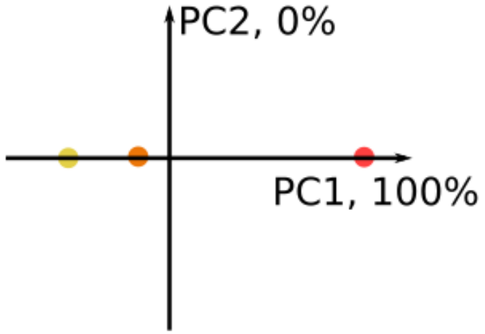
CDR3 3	5	1

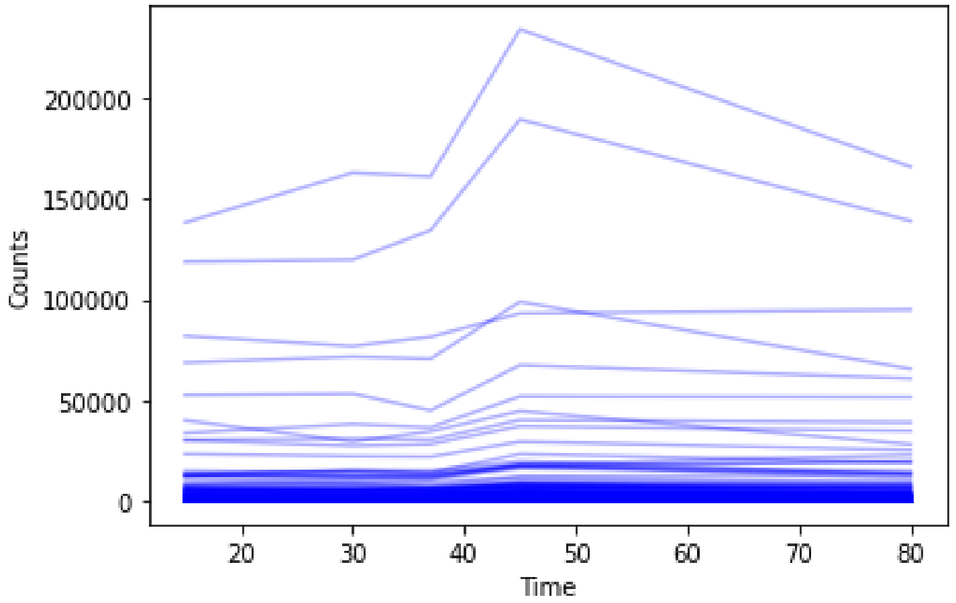






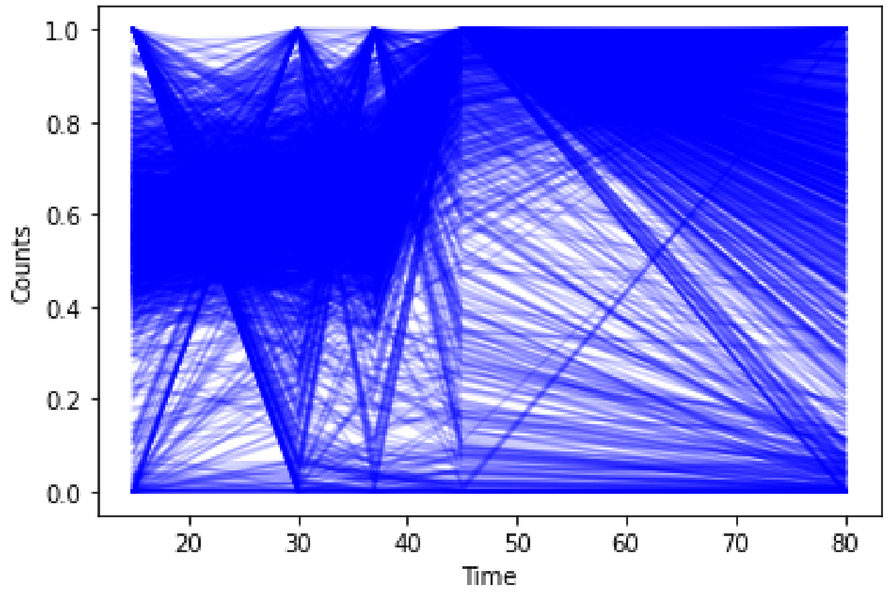
PCA for our 5d trajectories

We consider the CDR3 that have at least 1000 counts in at least 1 time point



PCA for our 5d trajectories

We consider the CDR3 that have at least 1000 counts in at least 1 time point and we normalize by the max count



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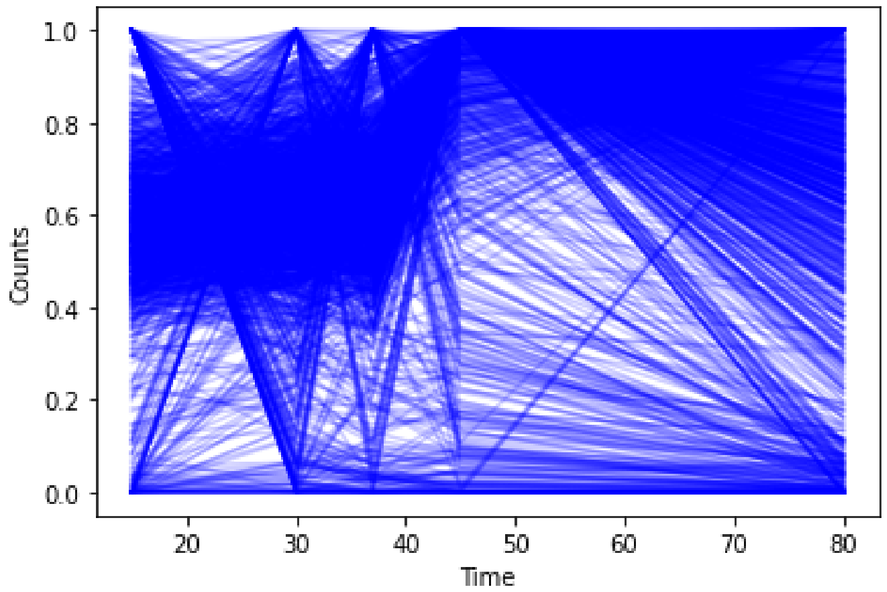

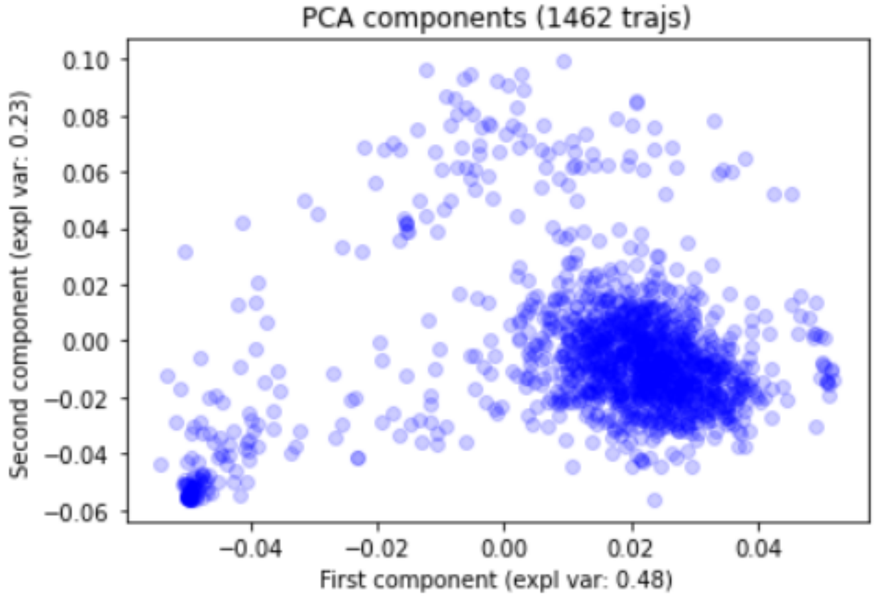

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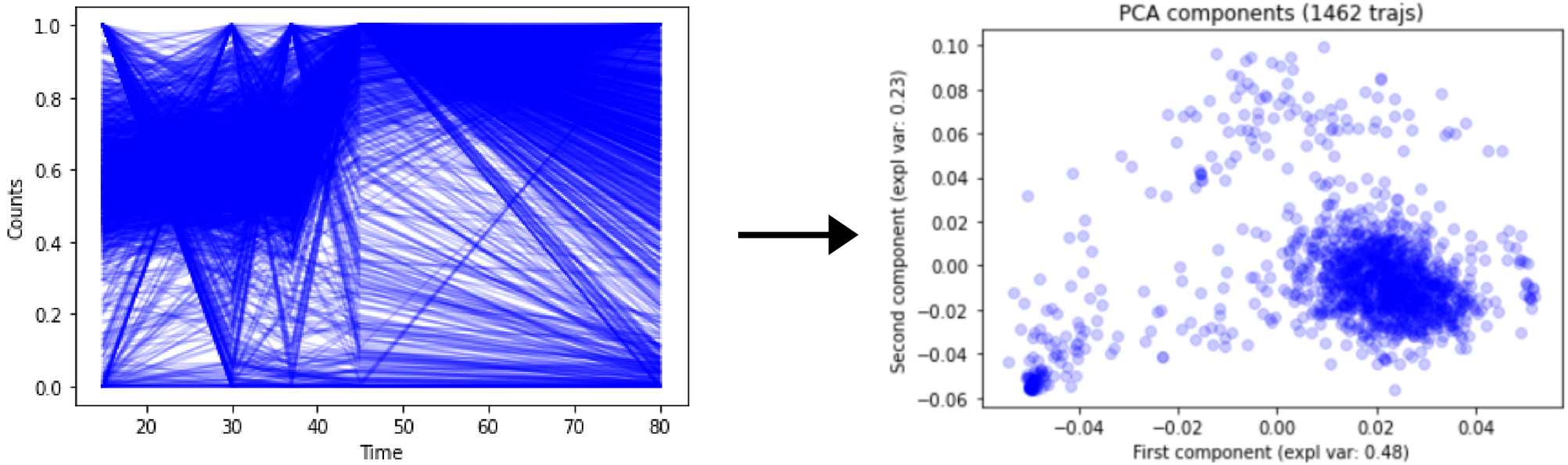

The best 2d representation of my 5d data. There is interesting structure!