

### Compatibility analysis of stat-of-the-art proposals with our proposal for K-means algorithm:

Paper	Contributions		Compatibility			
	Proposal	SIMD-computations considered?	SIMD-compression	Block-data-layout + optimized loop-traversal	In-register ArgMin computation	Approximate distance
[1]	Improved seeding algorithm	No	Yes	Yes	Yes	Yes
[2]	Heterogeneous computation: labeling on GPU, cluster update on CPU	Yes	Yes	Yes	Yes	Yes
PAMI02	Use of KD-tree to filter out a candidate.	No	Yes	No	No	Yes
[11]	Centroid oriented labeling, Heterogeneous computation: labeling on KNC, cluster update on CPU	Yes	Yes	No	No	Yes
Hamerly2010	Avoids distance computations using distance bounds and triangle inequality	No	Yes	No	No	Yes
Cui14	Use of MapReduce, iteration dependence is reduced using probability sampling	No	Yes	Yes	Yes	Yes
Wang13	Approximation using binary-tree based cluster closure	No	Yes	No	No	Yes
Wang15	Encode high dimensional data points	No	Yes	Yes	Yes	Yes

### Reference:

[Hamerly2010] G. Hamerly. Making k-means even faster. SMD'10.

[HiPC15] V-PFORDelta: Data Compression for Energy Efficient Computation of Time Series, HiPC'15.

[ARMSVE] <https://community.arm.com/groups/processors/blog/2016/08/22/technology-update-the-scalable-vector-extension-sve-for-the-armv8-a-architecture>

[Cui14] Cui et.al. Optimized big data K-means clustering using MapReduce. The Journal of Supercomputing, 2014.

[Wang13] Wang et.al. Fast approximate k-means via cluster closures, CVPR'12.

[Wang15] Wang et.al. Optimized Cartesian K-means, IEEE Transactions on Knowledge & Data Engineering, 2015.