

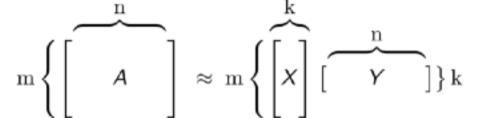
## Generalized Low-Rank Models

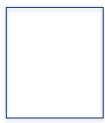
Principal Component Analysis (PCA). • Unlike PCA which is limited to numerical data, GLRM can also handle categorical, ordinal and Boolean data.

GLRM is an extension of well-known matrix factorization methods such as

- Given: Data table A with m rows and n columns • Find: Compressed representation as numeric tables X and Y where k is a small user-specified number

- Y = archetypal features created from columns of A
- X = row of A in reduced feature space
- GLRM can approximately reconstruct A from product XY















## Memory Reduction / Saving



## **H20 Generalized Low-Rank Models**

```
h2o.glrm(training_frame, cols = NULL, model_id = NULL, validation_frame = NULL,
    ignore_const_cols = TRUE, score_each_iteration = FALSE, loading_name = NULL,
    transform = c("NONE", "STANDARDIZE", "NORMALIZE", "DEMEAN", "DESCALE"), k = 1,
    loss = c("Quadratic", "Absolute", "Huber", "Poisson", "Hinge", "Logistic", "Periodic"),
    loss_by_col = c("Quadratic", "Absolute", "Huber", "Poisson", "Hinge", "Logistic",
    "Periodic", "Categorical", "Ordinal"), loss_by_col_idx = NULL,
    multi_loss = c("Categorical", "Ordinal"), period = 1,
    regularization_x = c("None", "Quadratic", "L2", "L1", "NonNegative", "OneSparse",
    "UnitOneSparse", "Simplex"), regularization_y = c("None", "Quadratic", "L2", "L1",
    "NonNegative", "OneSparse", "UnitOneSparse", "Simplex"), gamma_x = 0, gamma_y = 0,
    max_iterations = 1000, max_updates = 2000, init_step_size = 1,
    min_step_size = le-04, seed = -1, init = c("Random", "SVD", "PlusPlus", "User"),
    svd_method = c("GramSVD", "Power", "Randomized"), user_y = NULL, user_x = NULL,
    expand_user_y = TRUE, impute_original = FALSE, recover_svd = FALSE, max_runtime_secs = 0)
```

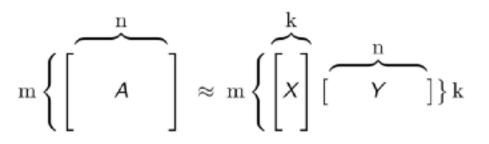
```
from h2o.estimators.glrm import H2OGeneralizedLowRankEstimator
model = H2OGeneralizedLowRankEstimator(...)
model.train(x = x, training_frame = data)
```





## Generalized Low-Rank Models

- GLRM is an extension of well-known matrix factorization methods such as Principal Component Analysis (PCA).
- Unlike PCA which is limited to numerical data, GLRM can also handle categorical, ordinal and Boolean data.
- Given: Data table A with m rows and n columns
- Find: Compressed representation as numeric tables X and Y where k is a small user-specified number



- Y = archetypal features created from columns of A
- *X* = row of A in reduced feature space
- GLRM can approximately reconstruct A from product XY

