



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

$$\begin{array}{c} \mathfrak{m} \end{array} \left\{ \begin{array}{c} \overbrace{\hspace{1.5cm}}^{\mathfrak{n}} \\ \left[\begin{array}{c} A \end{array} \right] \end{array} \right. \approx \begin{array}{c} \mathfrak{m} \end{array} \left\{ \begin{array}{c} \overbrace{\hspace{1.5cm}}^{\mathfrak{k}} \\ \left[\begin{array}{c} X \end{array} \right] \end{array} \right. \left[\begin{array}{c} \overbrace{\hspace{1.5cm}}^{\mathfrak{n}} \\ Y \end{array} \right] \} \mathfrak{k}$$











Memory Reduction / Saving



H2O Generalized Low-Rank Models

```
h2o.glm(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 = 1e-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.glm import H2OGeneralizedLowRankEstimator  
model = H2OGeneralizedLowRankEstimator(...)  
model.train(x = x, training_frame = data)
```



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$$m \left\{ \left[\begin{array}{c} \overbrace{\hspace{1cm}}^n \\ A \end{array} \right] \approx m \left\{ \left[\begin{array}{c} \overbrace{\hspace{1cm}}^k \\ X \end{array} \right] \left[\begin{array}{c} \overbrace{\hspace{1cm}}^n \\ Y \end{array} \right] \right\}^k$$

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