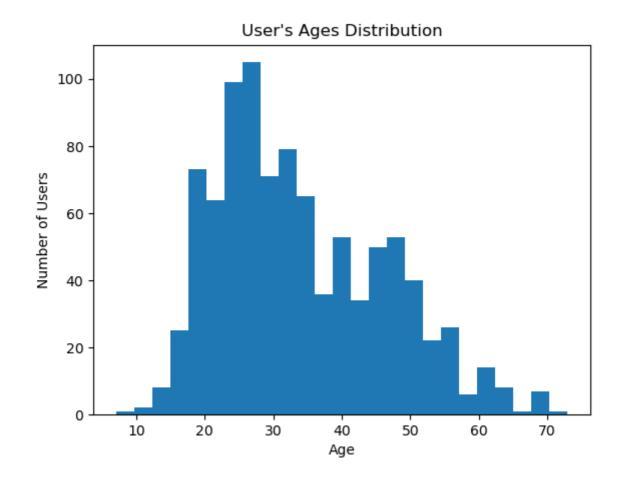
# Recommender System Project – Movielens100k

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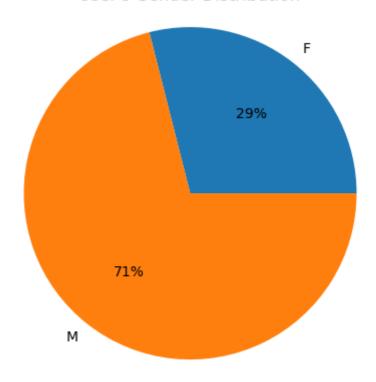
## Data discription

- 根據 ml-100k 中 readme.md 的描述:
   u.data 有 100000 筆資料, u1.base~u5.base 各有 80000 筆資料, u1.test~u5.test 各有 20000 筆資料。
- u.data u1~u5 中都有 4 個 columns:
   user id | item id | rating | timestamp
- u.user 中有 5 個 columns:user id | age | gender | occupation | zip code其中 age 的分佈爲:

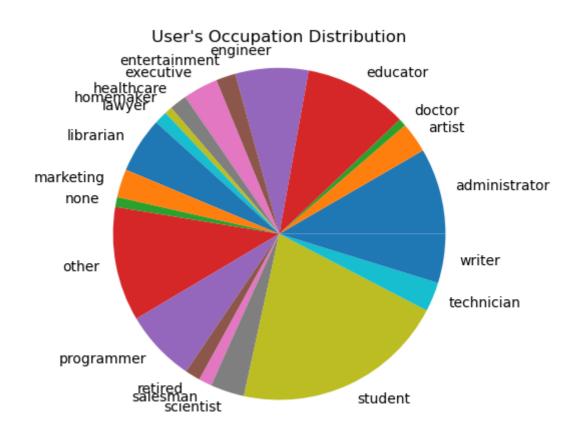


gender 的分佈為:

User's Gender Distribution



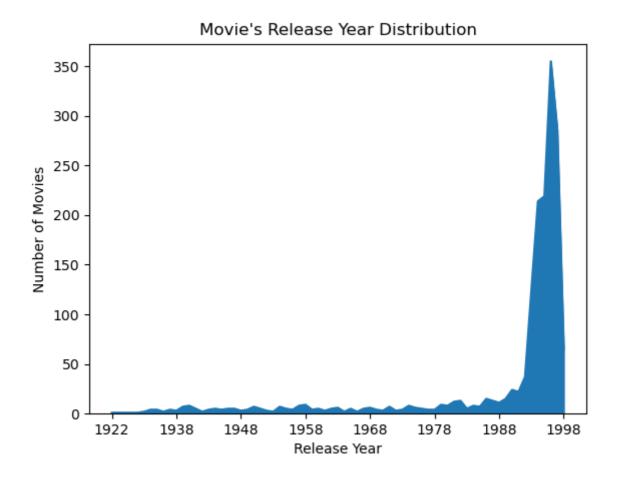
### occupation 的分佈為:



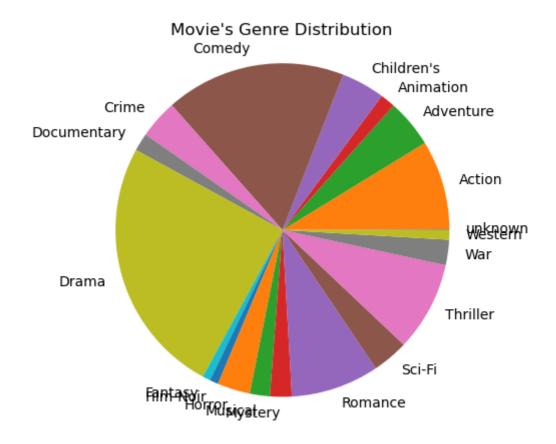
### • u.item 中有 24 個 columns:

movie id | movie title | release date | video release date | IMDb URL | unknown | Action | Adventure | Animation | Children's | Comedy | Crime | Documentary | Drama | Fantasy | Film-Noir | Horror | Musical | Mystery | Romance | Sci-Fi | Thriller | War | Western

其中 release date ( modified to release year ) 的分佈爲:



各個 genre 的分佈為:



### Program discription

Item Base Collaborative Filtering Model

### 實作方法:

- 1. 使用 pandas 的 pivot 將 rating data 轉換成 item-to-user 的稀疏矩陣。
- 2. 使用 sklearn 的 NearestNeighbors 由 item-to-user 矩陣取得 k-neighbor's cosine similarity 還有 k-neighbor's indices。
- 3. 照公式 \$\widehat{rate}=\cfrac{\sum (similarity \times rate)}{\sum similarity}\$ 得出預測的 rating 的值。
- 4. 計算 RMSE 衡量預測值的誤差。

#### Code 如下:

```
class Item_Base_CF:
    def __init__(self, n_neighbors=3) -> None:
        self.item_user = np.empty((1682, 943))
        self.n = n_neighbors
        self.knn = NearestNeighbors(n_neighbors=self.n, metric="cosine")
        self.distances, self.indices = np.empty((1682, self.n)),
        np.empty((1682, self.n))
        self.similarities = np.empty((1682, self.n))

def fit(self, data):
        self.item_user = data.pivot(index="item_id", columns="user_id",
```

```
values="rating")
        # Fail to adjust item-to-user matrix with mean of user-ratings
        # self.item user =
self.item user.subtract(self.item user.mean(axis=1), axis = 0).fillna(0)
        self.item user = self.item user.reindex(index=np.arange(1,1683),
fill value=0).fillna(0)
        # Use sklearn NearestNeighbors to find n similar item's indices &
cosine similarities
        self.knn.fit(self.item user)
        self.distances, self.indices =
self.knn.kneighbors(n neighbors=self.n)
        self.similarities = 1 - self.distances
    def predict(self, item_id, user id, epsilon=1e-8):
        # Function to predict rating via item id & user id
        # Using formula in 01 Neighborhood-
based collaborative filtering.pptx page 32
        pred r = self.item user.iloc[item id-1, user id-1]
        sim, ind = self.similarities[item id-1], self.indices[item id-1]
        product = 1
        product sum = 0
        if self.item user.iloc[item id-1, user id-1]!=0:
            return pred r
        else:
            for i in range(len(ind)):
                product = self.item user.iloc[ind[i],user id-1] * sim[i]
                product sum += product
            pred r = product sum / (np.sum(sim)+epsilon)
        return pred r
    def rmse(self, data):
        # Function to calculate RMSE of predicted rating & actual rating
        x = data[["item id", "user id"]].to numpy()
        y = data[["rating"]].to numpy()
        losses = []
        for i in range (len(x)):
            losses.append((self.predict(x[i][0], x[i][1]) - y[i]).item())
        return sqrt(np.nanmean(np.square(losses)))
```

### 輸出結果:

```
# Number of neighbors set k = [2, 4, 8]
```

使用 u1~u5 的資料集進行 5 次 RMSE 輸出,計算其平均值。

已知若使用所有預測值皆為 0 的 model, 其 5 次的 RMSE 輸出將會落在 3.6, 3.7 左右, 相較之下 Item Base CF 的誤差值有顯著的減少。

```
-----2-neighbors-----
u1.test RMSE : 2.8520523211052926
u2.test RMSE : 2.687808126654529
u3.test RMSE : 2.603465358260204
u4.test RMSE : 2.647911007490916
u5.test RMSE: 2.7360450020224687
mean : 2.705456363106682
----4-neighbors-----
ul.test RMSE : 2.7258851793259886
u2.test RMSE : 2.5596291495946533
u3.test RMSE : 2.4705296126894747
u4.test RMSE : 2.526707312903369
u5.test RMSE : 2.62215865725013
mean : 2.580981982352723
-----8-neighbors-----
ul.test RMSE : 2.6828780812337185
u2.test RMSE : 2.500036414134187
u3.test RMSE: 2.425303738135374
u4.test RMSE : 2.4748264591987574
u5.test RMSE: 2.575107201472646
mean : 2.5316303788349366
```