IP items recommender

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Keywords

- data mining
- ullet linear regression
- tags scores
- preferences prediction

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IPIRecModel 1

1.1 Series1

Algorithm 1 IPIRecModelSeries1

```
Require: No. of top n decisioned tags NT(u), threshold of co-occurred items
           Freq. \theta_I
  PREPROCESS(NT(u), \theta_I)
  PROCESS()
  function PREPROCESS(NT(u), \theta_I)
      top_n_decisioned_tags()
     mean_freq_tags()
  function PROCESS()
     item_based_tags_corr()
                                                                ▷ in Equation (1)
      user_based_tags_corr()
                                                                \triangleright in Equation (2)
                                                                ⊳ in Equation (3)
      tags_score()
```

$$sim_{U}(x,y) = \frac{\sum_{u \in U(x) \cap U(y)} (r(u,x) - \bar{r}(x)) (r(u,y) - \bar{r}(y))}{\sqrt{\sum_{u} (r(u,x) - \bar{r}(x))^{2} \sum_{u} (r(u,y) - \bar{r}(y))^{2}}}$$
(1)
$$sim_{I}(x,y) = \frac{\sum_{i \in I(x) \cap I(y)} (r(i,x) - \bar{r}(x)) (r(i,y) - \bar{r}(y))}{\sqrt{\sum_{i} (r(i,x) - \bar{r}(x))^{2} \sum_{i} (r(i,y) - \bar{r}(y))^{2}}}$$
(2)

$$sim_{I}(x,y) = \frac{\sum_{i \in I(x) \cap I(y)} (r(i,x) - \bar{r}(x)) (r(i,y) - \bar{r}(y))}{\sqrt{\sum_{i} (r(i,x) - \bar{r}(x))^{2} \sum_{i} (r(i,y) - \bar{r}(y))^{2}}}$$
(2)

$$s(x,y) = \text{CoOccur}(x,y) \cdot \Pr(x,y) \cdot \frac{1}{|\mathcal{S}|} \sum_{d \in \mathcal{S}} sim_d(x,y)$$
 (3)

$$\begin{split} \operatorname{CoOccur}(x,y) &= \frac{\ln(\min(|I(x)\cap I(y)|+1,\;\theta_I))}{\ln(\max_{x',y'\in T}|I(x')\cap I(y')|)};\;\theta_I = 20 \text{ (optional)} \\ &\operatorname{Pr}(x,y) = |I(x)|^{-1}|I(x)\cap I(y)| \\ &\bar{r}(x) = \frac{\sum_u |I(u)\cap I(V)|}{|U(x)\cap U(V)|} + \frac{\sum_i |U(i)\cap U(V)|}{|I(x)\cap I(V)|} \end{split}$$

1.2 Series2

Algorithm 2 IPIRecModelSeries2

```
Require: No. of top n decisioned tags NT(u), threshold of co-occurred items
           Freq. \theta_I
  PREPROCESS(NT(u), \theta_I)
  PROCESS()
  POSTPROCESS()
  function PREPROCESS(NT(u), \theta_I)
      top_n_decisioned_tags()
     mean_freq_tags()
  function PROCESS()
     item_based_tags_corr()
                                                                ▷ in Equation (1)
                                                                \triangleright in Equation (2)
      user_based_tags_corr()
                                                                \triangleright in Equation (3)
     tags_score()
  function POSTPROCESS()
                                                                ⊳ in Equation (4)
      append_tendencies()
```

$$\begin{split} S \leftarrow S + \mathbb{B}(\mathbb{X}(V), S), & (4) \\ \bar{b}(\mathbb{X}(V), x, y) = & \hat{b}^U(\mathbb{X}(V), x, y) + \hat{b}^I(\mathbb{X}(V), x, y) + \frac{b(\mathbb{X}(V), x) + b(\mathbb{X}(V), y)}{2} \\ & b(\mathbb{X}(V), t) = \frac{|R(t)|}{|R|} - \mu(\mathbb{X}(V)) \\ & \hat{b}^U(\mathbb{X}(V), x, y) = \frac{\sum_{u \in U(x) \cup U(y)} b(u)}{|U(x) \cup U(y)|} \\ & b(\mathbb{X}(V), u) = \frac{|I(u)|}{|I|} - \mu(\mathbb{X}(V)) \\ & \hat{b}^I(\mathbb{X}(V), x, y) = \frac{\sum_{i \in I(x) \cup I(y)} b(i)}{|I(x) \cup I(y)|} \\ & b(\mathbb{X}(V), i) = \frac{|U(i)|}{|U|} - \mu(\mathbb{X}(V)) \end{split}$$

1.3 Series3

Algorithm 3 IPIRecModelSeries3

```
Require: training set X, test set Y\rho = \operatorname{corr}(\mathbb{X}(V))\triangleright in Equation (5)\mathcal{B} = \operatorname{tendencies}(\mathbb{X}(V), \rho)\triangleright in Equation (6)S = \rho + \mathcal{B}\triangleright in Equation (7)
```

$$\rho(x,y) = J(x,y) \Pr(x,y) \prod_{d \in S} sim_d(x,y)$$

$$J(x,y) = \frac{|I(x) \cap I(y)|}{|I(x)|}$$

$$\Pr(x,y) = \frac{|I(x) \cap I(y)|}{|I(x)|}$$

$$sim_{UB}(x,y) = \frac{\sum_{u \in U(x) \cup U(y)} |NT(u) \cap x| |NT(u) \cap y|}{\sqrt{\sum_{u} |NT(u) \cap x| \sum_{u} |NT(u) \cap I(y)|}}$$

$$sim_{UF}(x,y) = \frac{\sum_{u \in U(x) \cup U(y)} |I(u) \cap I(x)| |I(u) \cap I(y)|}{\sqrt{\sum_{u} |I(u) \cap I(x)|^{2} \sum_{u} |I(u) \cap I(y)|^{2}}}$$

$$sim_{IB}(x,y) = \frac{\sum_{i \in I(x) \cup I(y)} |T(i) \cap x| |T(i) \cap y|}{\sqrt{\sum_{i} |T(i) \cap x|^{2} \sum_{i} |T(i) \cap y|^{2}}}$$

$$sim_{IF}(x,y) = \frac{\sum_{i \in I(x) \cup I(y)} \tau(x)\tau(y)}{\sqrt{\sum_{i} \tau(x)^{2} \sum_{i} \tau(y)^{2}}}$$

$$\tau(t) = |I(t)|^{-1} \sum_{i \in I(t)} |T(i)|$$

$$\bar{b}(x,y) = \hat{b}^{U}(x,y) + \hat{b}^{I}(x,y) + \frac{b(x) + b(y)}{2}$$

$$b(t) = \frac{|R(t)|}{|R|} - \mu(\mathbb{X})$$

$$\hat{b}^{U}(x,y) = \frac{\sum_{u \in U(x) \cup U(y)} b(u)}{|U(x) \cup U(y)|}$$

$$b(u) = \frac{|I(u)|}{|I|} - \mu(\mathbb{X})$$

$$\hat{b}^{I}(x,y) = \frac{\sum_{i \in I(x) \cup I(y)} b(i)}{|I(x) \cup I(y)|}$$

$$b(i) = \frac{|U(i)|}{|U|} - \mu(\mathbb{X})$$

$$s(x,y) \leftarrow \rho(x,y) + \bar{b}(x,y)$$
(7)

2 Estimators

2.1 Series1

Algorithm 4 IPIRecEstimatorSeries1

```
Require: iter
   ADJUST\_TAGS\_SCORE(X(V))
   for a \in [V, L, P] do
       for _{-} in range(iter) do
            TRAINING(\mathbb{X}(a))
   function ADJUST_TAGS_SCORE(X)
       W = \text{numpy.ones}(|T|, |T|)
        for r(u,i) \in \mathbb{X} do
            \hat{r}(u,i) = \text{estimate}(u,i)
                                                                                   \triangleright in Equation (8)
            W = \text{feed\_back}(r(u, i), \hat{r}(u, i), W)
                                                                                   \triangleright in Equation (9)
       \overline{S} = SW
   function Training(X)
        for r(u,i) \in \mathbb{X} do
            \hat{r}(u,i) = \text{estimate}(u,i)
                                                                                   \triangleright in Equation (8)
            \Omega = \text{feed\_back}(r(u, i), \hat{r}(u, i), \Omega)
                                                                                  \triangleright in Equation (10)
```

$$\hat{r}(u,i) = \frac{\sum_{x} \sum_{y} \omega(u,x,y) s(x,y) \nu(u,y)}{\sum_{x \in NT(u)} \sum_{y \in T(i)} |\omega(u,x,y)|}$$
(8)
$$\nu(u,y) = \begin{cases} NT(u) \cap y = \phi & \text{Default voting} \\ \text{otherwise} & 1 \end{cases}$$

$$\epsilon(u,i) = r(u,i) - \hat{r}(u,i)$$

$$w(x,y) = \frac{w + \eta \left(w(x,y) \epsilon(u,i) \lambda^{S} \gamma^{S} s(x,y)\right)}{\sum_{z \in T(i)} |w(x,z) s(x,z)|}$$
(9)
$$\eta(x) = \begin{cases} x \geq 0 & 2^{-1} x \\ \text{otherwise} & 2x \end{cases}$$

$$\mathbb{W}(u) = \sum_{x \in NT(u)} \sum_{y \in T(i)} |\omega(u,x,y)|$$

$$\omega(u,x,y) = \omega(u,x,y) \frac{r(u,i) + \lambda^{\Omega} s(x,y)}{\hat{r}(u,i) + \gamma^{\Omega} \mathbb{W}(u)}$$
(10)

2.2 Series2

Algorithm 5 IPIRecEstimatorSeries2

```
Require: iter
  \Omega = \text{numpy.ones}(|U|, |T|, |T|)
  ADJUST\_TAGS\_SCORE(X(V))
  for a \in [V, L, P] do
       for _{-} in range(iter) do
           TRAINING(\mathbb{X}(a))
  function Adjust_tags_score(X)
       W = \text{numpy.ones}(|T|, |T|)
       for r(u,i) \in \mathbb{X} do
           for x \in NT(u) do
                for y \in T(i) do
                    \hat{r}(u,i) = \text{estimate}(u,i)
                                                                             ▷ in Equation (11)
                    W = \text{feed\_back}(r(u, i), \hat{r}(u, i), W)
                                                                             \triangleright in Equation (12)
       \bar{S} = S\bar{W}
  function Training(X)
       for r(u,i) \in \mathbb{X} do
                                                                             ▷ in Equation (11)
           \hat{r}(u,i) = \text{estimate}(u,i)
           \Omega = \text{fit\_weights}(r(u, i), \hat{r}(u, i), S, \Omega)
                                                                             \triangleright in Equation (13)
           S = \text{fit\_scores}(r(u, i), \hat{r}(u, i), S, \Omega)
                                                                             ▷ in Equation (14)
```

$$\hat{r}(u,i) = \frac{\sum_{x} \sum_{y} \omega(u,x,y) s(x,y) \nu(u,y)}{\sum_{x \in NT(u)} \sum_{y \in T(i)} |\omega(u,x,y)|}$$

$$\nu(u,y) = \begin{cases} NT(u) \cap y = \phi & DV \\ \mathbf{otherwise} & 1 \end{cases}$$

$$\epsilon(u,i) = r(u,i) - \hat{r}(u,i)$$

$$(11)$$

$$s(x,y) = s(x,y) + \lambda^{S} \eta \left(\epsilon(u,i) \left(\frac{s(x,y)}{\|S\|_{2}} \right) \right) + \gamma^{S} \frac{\bar{s}(x) + \bar{s}(y)}{2}$$

$$\eta(x) = \begin{cases} x \ge 0, 2^{-1}x \\ \text{otherwise}, 2x \end{cases}$$

$$(12)$$

$$\omega(u, x, y) = \omega(u, x, y) \frac{(r(u, i) + \alpha(\hat{r}(u, i))) + \lambda^{\Omega} s(x, y)}{(\hat{r}(u, i) + \alpha(\hat{r}(u, i))) + \gamma^{\Omega} \mathbb{W}(u, i)}$$

$$\alpha(x) = \begin{cases} x = 0 & 1 \\ \text{otherwise} & 0 \end{cases}$$
(13)

$$\mathbb{W}(u,i) = \sum_{x \in NT(u)} \sum_{y \in T(i)} |\omega(u,x,y)|$$
$$s(x,y) = s(x,y) \frac{(r(u,i) + \alpha(\hat{r}(u,i))) + \lambda^{\Omega}\omega(u,x,y)}{(\hat{r}(u,i) + \alpha(\hat{r}(u,i))) + \gamma^{\Omega}\mathbb{S}(u,i)}$$

(14)

$$\mathbb{S}(u,i) = \sum_{NT(u)} \sum_{y \in T(i)} |s(x,y)|$$

2.3 Series3

Algorithm 6 IPIRecEstimatorSeries3

```
Require: Train.Set. \mathbb{X}, main Seq. \mathcal{A}^{\text{main}}, post Seq. \mathcal{A}^{\text{post}}
    \Omega \leftarrow \text{np.ones(shape}=(|U|,|T|,|T|))
    S = \text{Adjust\_scores}(\mathbb{X}(V), S, \Omega)
    for a \in \mathcal{A}^{\min} do
           \mathcal{L}_{\min} = \infty
           while True do
                 \mathbb{L}^{\Omega}, S, \Omega = \text{Fit_Weights}(\mathbb{X}, a, S, \Omega)
                 \mathbb{L}^S, S, \Omega = \text{FIT\_SCORES}(\mathbb{X}, a, S, \Omega)
                 \mathcal{L} = \frac{2\mathbb{L}^{S}\mathbb{L}^{\Omega}}{\mathbb{L}^{S} + \mathbb{L}^{\Omega}}
                                                                                                                   \triangleright in Equation (18)
                 if \mathcal{L}_{\min} \stackrel{\mathbb{L}^{\mathcal{S}} + \mathbb{L}^{\mathcal{S}}}{\geq} \mathcal{L} then
                        \mathcal{L}_{\min} = \mathcal{L}
                        S_{\min} = S
                       \Omega_{\min} = \Omega
                 else
                        \mathcal{L} = \mathcal{L}_{\min}
                        S = S_{\min}
                        \Omega = \Omega_{\min}
                       break
    for a \in \mathcal{A}^{\mathrm{post}} do
          if a \neq V then
           \_, S, \Omega = \text{FIT\_SCORES}(\mathbb{X}(V), S, \Omega)
           _{-}, S, \Omega = \text{Fit\_scores}(\mathbb{X}(a), S, \Omega)
    function ADJUST_SCORES(X, S, \Omega)
           S, L = FIT\_SCORES(X, V, S, \Omega)
          \mathcal{B} = \text{tendencies}(\mathbb{X}, S)
                                                                                                                     \triangleright in Equation (6)
          S = S + \mathcal{B}
           S, S, S = FIT\_SCORES(X, V, S, \Omega)
          return S
    function FIT_SCORES(\mathbb{X}, a, S, \Omega)
          if a \neq V then
                 S, \Omega = \arg_{S',\Omega'} \min \mathbb{L}(\mathbb{X}(V), S, \Omega)
          return \arg_{\mathbb{L}',S',\Omega'} \min \mathbb{L}(\mathbb{X}(a),S,\Omega)
    function FIT_WEIGHTS(\mathbb{X}, a, S, \Omega)
          \mathbb{L}, S, \Omega = \arg_{\mathbb{L}', S', \Omega'} \min \mathbb{L}(\mathbb{X}(a), S, \Omega)
           , S, \Omega = \text{FIT\_SCORES}(\mathbb{X}, a, S, \Omega)
          return \mathbb{L}, S, \Omega
```

$$\hat{r}(u,i) = \frac{1}{|T(u)|} \sum_{x \in T(u)} \frac{\prod_{y \in T(i)} \omega(u,x,y) s(x,y)}{\prod_{y \in T(i)} |\omega(u,x,y)|}$$
(15)
$$\epsilon(u,i) = r(u,i) - \hat{r}(u,i)$$

$$s(x,y) = s(x,y) + \lambda^{S} \eta \left(\epsilon(u,i) \left(\frac{s(x,y)}{\|S\|_{2}} + \frac{\bar{s}(x) + \bar{s}(y)}{2} \right) \right)$$
(16)
$$\eta(x) = \begin{cases} x \geq 0 & 2^{-1} x \\ \text{otherwise } 2x \end{cases}$$

$$\omega(u,x,y) = \omega(u,x,y) \frac{r(u,i) + \lambda^{\Omega} s(x,y)}{\hat{r}(u,i) + \gamma^{\Omega} \mathbb{W}(u,i)}$$

$$\mathbb{W}(u,i) = \sum_{x \in NT(u)} \sum_{y \in T(i)} |\omega(u,x,y)|$$

$$s(x,y) = s(x,y) \frac{r(u,i) + \lambda^{\Omega} \omega(u,x,y)}{\hat{r}(u,i) + \gamma^{\Omega} \mathbb{S}(u,i)}$$

$$\mathbb{S}(u,i) = \sum_{NT(u)} \sum_{y \in T(i)} |s(x,y)|$$

$$\mathbb{L}(\mathbb{X}, S, \Omega) = \text{RMSE}(\mathbb{X}, S, \Omega) = \sqrt{\frac{1}{|\mathbb{X}|} \sum_{(u,i) \in \mathbb{X}} (r(u,i) - \hat{r}(u,i))^{2}}$$

$$\mathbb{L}(\mathbb{X}, S, \Omega) = \min_{\mathbf{R}} \text{RMSE}^{S}(\mathbb{X}(a), S, \Omega)$$

$$\mathbb{L}^{S}(\mathbb{X}(a), S, \Omega) = \mathbb{L}^{S}(\mathbb{X}(a), S, \Omega) + \mathbb{L}^{S}(\mathbb{X}(a), S, \Omega)$$

$$\min_{\mathbf{L}} \mathcal{L}(\mathbb{X}, S, \Omega)$$

$$\min_{\mathbf{L}} \mathcal{L}(\mathbb{X}, S, \Omega)$$

$$\min_{\mathbf{L}} \mathcal{L}(\mathbb{X}, S, \Omega)$$
(18)

2.4 Series4

식 (19)를 사용한 알고리즘 6으로 모델을 훈련함

$$\mathcal{F}(\mathbb{X}(a), S, \Omega) = \frac{|\mathbb{X}(a)|}{|U|^2 |I|} \sum_{u \in U} \left(\frac{1}{|T(u)||T|} \sum_{x \in T(u)} \sum_{y \in T} \left(w(u, x, y) s(x, y) \right)^2 \right)^{0.5}$$

$$\mathbb{L}(\mathbb{X}, S, \Omega) = \text{RMSE}(\mathbb{X}, S, \Omega) + \mathcal{F}(\mathbb{X}(a), S, \Omega)$$
(19)