
Algorithm 1 Explicit Shape Regression (ESR)

Variables: Training images and labeled shapes $\{I_l, \hat{S}_l\}_{l=1}^L$; ESR model $\{R^t\}_{t=1}^T$; Testing image I ; Predicted shape S ; *TrainParams*{times of data augment N_{aug} , number of stages T }; *TestParams*{number of multiple initializations N_{int} }; *InitSet* which contains exemplar shapes for initialization

ESRTraining($\{I_l, \hat{S}_l\}_{l=1}^L, \text{TrainParams}, \text{InitSet}$)
// augment training data
 $\{I_i, \hat{S}_i, S_i^0\}_{i=1}^N \leftarrow \text{Initialization}(\{I_l, \hat{S}_l\}_{l=1}^L, N_{\text{aug}}, \text{InitSet})$
for t from 1 to T
 $Y \leftarrow \{M_{S_i^{t-1}} \circ (\hat{S}_i - S_i^{t-1})\}_{i=1}^N$ // compute normalized targets
 $R^t \leftarrow \text{LearnStageRegressor}(Y, \{I_i, S_i^{t-1}\}_{i=1}^N)$ // using Eq. (3)
 for i from 1 to N
 $S_i^t \leftarrow S_i^{t-1} + M_{S_i^{t-1}}^{-1} \circ R^t(I_i, S_i^{t-1})$
return $\{R^t\}_{t=1}^T$

ESRTesting($I, \{R^t\}_{t=1}^T, \text{TestParams}, \text{InitSet}$)
// multiple initializations
 $\{I_i, *, S_i^0\}_{i=1}^{N_{\text{int}}} \leftarrow \text{Initialization}(\{I, *\}, N_{\text{int}}, \text{InitSet})$
for t from 1 to T
 for i from 1 to N_{int}
 $S_i^t \leftarrow S_i^{t-1} + M_{S_i^{t-1}}^{-1} \circ R^t(I_i, S_i^{t-1})$
 $S \leftarrow \text{CombineMultipleResults}(\{S_i^T\}_{i=1}^{N_{\text{int}}})$
return S

Initialization($\{I_c, \hat{S}_c\}_{c=1}^C, D, \text{InitSet}$)

$i \leftarrow 1$
for c from 1 to C
 for d from 1 to D
 $S_i^0 \leftarrow$ sampling an exemplar shape from *InitSet*
 $\{I_i^0, \hat{S}_i^0\} \leftarrow \{I_c, \hat{S}_c\}$
 $i \leftarrow i + 1$
return $\{I_i^0, \hat{S}_i^0, S_i^0\}_{i=1}^{CD}$

Algorithm 3 Correlation-based feature selection

Input: regression targets $Y \in \mathbb{R}^{N \times 2N_{\text{fp}}}$; shape indexed pixel features $\rho \in \mathbb{R}^{N \times P}$; pixel-pixel covariance $\text{cov}(\rho) \in \mathbb{R}^{P \times P}$; number of features of a fern F ;

Output: The selected pixel-difference features $\{\rho_{m_f} - \rho_{n_f}\}_{f=1}^F$ and the corresponding indices $\{m_f, n_f\}_{f=1}^F$;

CorrelationBasedFeatureSelection($Y, \text{cov}(\rho), F$)

for f from 1 to F
 $v \leftarrow \text{randn}(2N_{\text{fp}}, 1)$ // draw a random projection from unit Gaussian
 $Y_{\text{prob}} \leftarrow Yv$ // random projection
 $\text{cov}(Y_{\text{prob}}, \rho) \in \mathbb{R}^{1 \times P} \leftarrow$ compute target-pixel covariance
 $\sigma(Y_{\text{prob}}) \leftarrow$ compute sample variance of Y_{prob}
 $m_f = 1; n_f = 1$;
 for m from 1 to P
 for n from 1 to P
 $\text{corr}(Y_{\text{prob}}, \rho_m - \rho_n) \leftarrow$ compute correlation using Eq. (11)
 if $\text{corr}(Y_{\text{prob}}, \rho_m - \rho_n) > \text{corr}(Y_{\text{prob}}, \rho_{m_f} - \rho_{n_f})$
 $m_f = m; n_f = n$;
return $\{\rho_{m_f} - \rho_{n_f}\}_{f=1}^F, \{m_f, n_f\}_{f=1}^F$

Algorithm 2 Shape indexed features

Variables: images and corresponding estimated shapes $\{I_i, S_i\}_{i=1}^N$; number of shape indexed pixel features P ; number of facial points N_{fp} ; the range of local coordinate κ ; local coordinates $\{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P$; shape indexed pixel features $\rho \in \mathbb{R}^{N \times P}$; shape indexed pixel-difference features $X \in \mathbb{R}^{N \times P^2}$;

GenerateShapeIndexedFeatures($\{I_i, S_i\}_{i=1}^N, N_{\text{fp}}, P, \kappa$)
 $\{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P \leftarrow \text{GenerateLocalCoordinates}(\text{FeatureParams})$
 $\rho \leftarrow \text{ExtractShapeIndexedPixels}(\{I_i, S_i\}_{i=1}^N, \{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P)$
 $X \leftarrow$ pairwise difference of all columns of ρ
return $\{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P, \rho, X$

GenerateLocalCoordinates(N_{fp}, P, κ)

for α from 1 to P
 $l_{\alpha} \leftarrow$ randomly drawn a integer in $[1, N_{\text{fp}}]$
 $\Delta_{\alpha}^{l_{\alpha}} \leftarrow$ randomly drawn two floats in $[-\kappa, \kappa]$
return $\{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P$

ExtractShapeIndexedPixels($\{I_i, S_i\}_{i=1}^N, \{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P$)

for i from 1 to N
 for α from 1 to P
 $\mu_{\alpha} \leftarrow \pi_{l_{\alpha}} \circ S_i + M_{S_i}^{-1} \circ \Delta_{\alpha}^{l_{\alpha}}$
 $\rho_{i\alpha} \leftarrow I_i(\mu_{\alpha})$
return ρ

Algorithm 4 Internal-level boosted regression

Variables: regression targets $Y \in \mathbb{R}^{N \times 2N_{\text{fp}}}$; training images and corresponding estimated shapes $\{I_i, S_i\}_{i=1}^N$; training parameters *TrainParams*{ $N_{\text{fp}}, P, \kappa, F, K$ }; the stage regressor R ; testing image and corresponding estimated shape $\{I, S\}$;

LearnStageRegressor($Y, \{I_i, S_i\}_{i=1}^N, \text{TrainParams}$)
 $\{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P \leftarrow \text{GenerateLocalCoordinates}(N_{\text{fp}}, P, \kappa)$
 $\rho \leftarrow \text{ExtractShapeIndexedPixels}(\{I_i, S_i\}_{i=1}^N, \{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P)$
 $\text{cov}(\rho) \leftarrow$ pre-compute pixel-pixel covariance
 $Y^0 \leftarrow Y$ // initialization
for k from 1 to K
 $\{\rho_{m_f} - \rho_{n_f}\}_{f=1}^F, \{m_f, n_f\}_{f=1}^F \leftarrow$
 CorrelationBasedFeatureSelection($Y^{k-1}, \text{cov}(\rho), F$)
 $\{\theta_f\}_{f=1}^F \leftarrow$ sample F thresholds from an uniform distribution
 $\{\Omega_b\}_{b=1}^{2^F} \leftarrow$ partition training samples into 2^F bins
 $\{y_b\}_{b=1}^{2^F} \leftarrow$ compute the outputs of all bins using Eq. (7)
 $r_k \leftarrow \{\{m_f, n_f\}_{f=1}^F, \{\theta_f\}_{f=1}^F, \{y_b\}_{b=1}^{2^F}\}$ // construct a fern
 $Y^k \leftarrow Y^{k-1} - r^k(\{\rho_{m_f} - \rho_{n_f}\}_{f=1}^F)$ // update the targets
 $R \leftarrow \{r^k\}_{k=1}^K, \{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P$ // construct stage regressor
return R

ApplyStageRegressor(I, S, R) // i.e. $R(I, S)$
 $\rho \leftarrow \text{ExtractShapeIndexedPixels}(\{I, S\}, \{\Delta_{\alpha}^{l_{\alpha}}\}_{\alpha=1}^P)$
 $\delta S \leftarrow \mathbf{0}$
for k from 1 to K
 $\delta S \leftarrow \delta S + r^k(\{\rho_{m_f} - \rho_{n_f}\}_{f=1}^F)$
return δS
