**Supplementary information**

Jieyun Bai1,2,\*, Zhanhang Sun1, Sheng Yu1, Yaosheng Lu1,2, Shun Long1, Huijin Wang1, Ruiyu Qiu1, Zhanhong Ou1, Minghong Zhou1, Dengjiang Zhi1, Mengqiang Zhou1, Xiaosong Jiang1, and Gaowen Chen3\*

1College of Information Science and Technology, Jinan University, Guangzhou 510632, China;

2Guangdong Provincial Key Laboratory of Traditional Chinese Medicine Information Technology, Jinan University, Guangzhou 510632, China;

3Obstetrics and Gynecology Center, Zhujiang Hospital, Southern Medical University, Guangzhou 510280, China.

Correspondence and requests for materials should be addressed to Jieyun Bai (baijieyun@jnu.edu.cn or bai\_jieyun@126.com) or Gaowen Chen (cgw2012@163. com)

**Supplemental Table 1.** The configurations of the DBSN.

|  |  |  |  |
| --- | --- | --- | --- |
| Layer | Input | Operator | Output |
| *F1* | Original image | (Conv+GN+ReLu)×2 | En\_Map1-0(512×384×32) |
| En\_Map1-0 | 2×2 Max Pooling | En\_Map1-1(256×192×32) |
| *F2* | En\_Map1-1 | (Conv+GN+ReLu)×2 | En\_Map2-0(256×192×64) |
| En\_Map2-0 | 2×2 Max Pooling | En\_Map2-1(128×96×64) |
| *F3* | En\_Map2-1 | (Conv+GN+ReLu)×2 | En\_Map3-0(128×96×128) |
| En\_Map3-0 | 2×2 Max Pooling | En\_Map3-1(64×48×128) |
| *F4* | En\_Map3-1 | (Conv+GN+ReLu)×2 | En\_Map4-0(64×48×256) |
| En\_Map4-0 | 2×2 Max Pooling | En\_Map4-1(32×24×256) |
| *F5* | En\_Map4-1 | (Conv+GN+ReLu)×2 | En\_Map5-1(32×24×512) |
| *L1* | En\_Map5-1 | (Conv+GN+ReLu)×2 | De\_L\_Map1-1(32×24×256) |
| *L2* | De\_L\_Map1-1 | (Conv+GN+ReLu)×2 | De\_L\_Map2-1(32×24×128) |
| *L3* | De\_L\_Map2-1 | (Conv+GN+ReLu)×2 | De\_L\_Map3-1(32×24×64) |
| *L4* | De\_L\_Map3-1 | (Conv+GN+ReLu)×2 | De\_L\_Map4-1(32×24×32) |
| *DL* | De\_L\_Map4-1 | 1×1Conv + Softmax | Lower\_Output:(32×24×3) |
| *U1* | De\_L\_Map1-1 | Up-sampling+Conv+GN+ ReLu | De\_L\_U1-1(64×48×256) |
| *U2* | En\_Map4-0De\_L\_U1-1 | (Conv+GN+ReLu)×2 | De\_U\_Map2-1(64×48×128) |
| De\_L\_Map2-1 | Up-sampling+Conv+GN+ ReLu | De\_L\_U2-1(128×96×128) |
| De\_U\_Map2-1 | Up-sampling+DC+GN+ ReLu | De\_U\_U2-1(128×96×128) |
| De\_L\_U2-1De\_U\_U2-1 | Attention Gate | De\_L\_U AG2-1(128×96×128) |
| *U3* | En\_Map3-0De\_U\_U2-1De\_L\_U AG2-1 | (Conv+GN+ReLu)×2 | De\_U\_Map3-0(128×96×64) |
| De\_L\_Map3-1 | Up-sampling+Conv+GN+ ReLu | De\_L\_U3-1256×192×64 |
| De\_U\_Map3-0 | Up-sampling+ DC+GN+Relu | De\_U\_U3-1(256×192×64) |
| De\_L\_U3-1De\_U\_U3-1 | Attention Gate | De\_L\_U AG3-1(256×192×64) |
| En\_Map2-0De\_U\_U3-1De\_L\_U AG3-1 | (Conv+GN+ReLu)×2 | De\_U\_Map3-1(256×192×64) |
| *U4* | De\_L\_Map4-1 | Up-sampling+Conv+GN+ReLu | De\_L\_U4-1(512×384×32) |
| De\_U\_Map3-1 | Up-sampling+DC+GN+Relu | De\_U\_U4-1(512×384×32) |
| De\_L\_U4-1De\_U\_U4-1 | Attention Gate | De\_L\_U AG4-1(512×384×32) |
| En\_Map1-0De\_U\_U4-1De\_L\_U AG4-1 | (Conv+GN+ReLu)×2 | De\_U\_Map4-1512×384×32 |
| *DU* | De\_U\_Map4-1 | 1×1Conv + Softmax | Upper\_Output:(512×384×3) |

Note: Conv: Convolution operator; DC: Deformable Convolution operator; GN: Group normalization.

**Supplemental Table 2.** Performance of the network using different weights (***wL***) on Acc, ***Diceall***, ***DicePS*** and ***DiceFH*** on our private dataset. (The blacked part denoted those indicators are the best.)

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| ***wL*** | ***Acc*** | ***Diceall*** | ***DicePS*** | ***DiceFH*** |
| 0.1 | 0.9865 | 0.9291 | 0.9005 | 0.9327 |
| **0.2** | **0.9874** | **0.9338** | **0.9101** | **0.9366** |
| 0.3 | 0.9867 | 0.9301 | 0.9096 | 0.9326 |
| 0.5 | 0.9868 | 0.9310 | 0.9059 | 0.9341 |
| 1.0 | 0.9867 | 0.9298 | 0.9040 | 0.9330 |

Note: ***wL*** denotes the weight of the loss function for the lower branch of DBSN.

**Supplemental Table 3.** Performance of different networks on Acc, ***Diceall***, ***DicePS*** and ***DiceFH*** on our private dataset. (The blacked part denoted those indicators are the best.)

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| ***Model*** | ***Acc*** | ***Diceall*** | ***DicePS*** | ***DiceFH*** |
| *DBSN* | **0.9874** | 0.9338 | 0.9101 | 0.9366 |
| *DBSN-AG* | **0.9874** | **0.9339** | 0.9084 | **0.9369** |
| *DBSN-LB* | 0.9873 | 0.9334 | **0.9156** | 0.9355 |
| *DBSN-DC* | 0.9864 | 0.9292 | 0.9083 | 0.9317 |

Note: DBSN, DBSN-AG, DBSN-LB and DBSN-DC denotes, Double Branch Segmentation Network (DBSN), DBSN without the attention gate (AG), DBSN without the low branch (LB) and DBSN without deformable convolutional blocks (DC), respectively.

**Supplemental Table 4.** Performance of different networks on ***DU***, ***DL*** and ***AX*** on our private dataset. (The blacked part denoted those indicators are the best.)

|  |  |  |  |
| --- | --- | --- | --- |
| ***Model*** | ***EDU* (mm)** | ***EDL* (mm)** | ***AX* (**°**)** |
| *DBSN* | **4.21** | 3.62 | 4.15 |
| *DBSN-AG* | 4.51 | 3.67 | 4.48 |
| *DBSN-LB* | 4.47 | **3.52** | **3.97** |
| *DBSN-DC* | 4.69 | 3.84 | 4.30 |

Note: DBSN, DBSN-AG, DBSN-LB and DBSN-DC denotes, Double Branch Segmentation Network (DBSN), DBSN without the attention gate (AG), DBSN without the low branch (LB) and DBSN without deformable convolutional blocks (DC), respectively.

**Supplemental Table 5.** Performance of different networks on $∆$*AoP\_Mean*, $∆$*AoP\_Median* and $∆$*AoP\_Std*on our private dataset. (The blacked part denoted those indicators are the best.)

|  |  |  |  |
| --- | --- | --- | --- |
| ***Model*** | $∆$***AoP\_Mean* (°)** | $∆$***AoP\_Median* (°)** | $∆$***AoP\_Std* (°)** |
| *DBSN* | **5.993** | 5.851 | **3.872** |
| *DBSN-AG* | 6.490 | 5.690 | 5.151 |
| *DBSN-LB* | 6.200 | **5.005** | 5.096 |
| *DBSN-DC* | 6.456 | 5.180 | 6.043 |

Note: DBSN, DBSN-AG, DBSN-LB and DBSN-DC denotes, Double Branch Segmentation Network (DBSN), DBSN without the attention gate (AG), DBSN without the low branch (LB) and DBSN without deformable convolutional blocks (DC), respectively.

**Supplemental Table 6.** Performance of different networks on $∆$*AoP\_Mean*, $∆$*AoP\_Median* and $∆$*AoP\_Std*on the public JNU-IFM dataset. (The blacked part denoted those indicators are the best.)

|  |  |  |  |
| --- | --- | --- | --- |
| ***Model*** | ***Diceall*** | ***DicePS*** | ***DiceFH*** |
| *DBSN* | 0.9180 | 0.8726 | 0.9234 |
| *DBSN-AG* | **0.9198** | 0.8740 | **0.9253** |
| *DBSN-LB* | 0.9188 | **0.8760** | 0.9240 |
| *DBSN-DC* | 0.9152 | 0.8696 | 0.9208 |

**Supplemental Table 7.** Performance of different networks on $∆$***AoP\_Mean***, $∆$***AoP\_Median***and $∆$***AoP\_Std*** on the public JNU-IFM dataset. (The blacked part denoted those indicators are the best.)

|  |  |  |  |
| --- | --- | --- | --- |
| ***Model*** | $∆$***AoP\_Mean* (°)** | $∆$***AoP\_Median* (°)** | $∆$***AoP\_Std* (°)** |
| *DBSN* | 5.178 | 4.229 | 4.579 |
| *DBSN-AG* | 5.188 | 4.260 | 4.542 |
| *DBSN-LB* | 5.133 | 4.211 | **4.338** |
| *DBSN-DC* | **5.110** | **4.181** | 4.607 |

**Supplemental Table 8.** Performance of U-net and DBSN on our private dataset.

|  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- |
| ***Model*** | ***ACC*** | ***Diceall*** | ***DicePS*** | ***DiceFH*** | ***ASD*** | $∆$***AoP\_Mean* (°)** |
| U-net | 0.984 | 0.913 | 0.889 | 0.916 | 10.989 | 8.00 |
| **DBSN** | **0.987** | **0.934** | **0.910** | **0.937** | **6.268** | **5.993** |



**Supplemental Fig. 1**. Example feature maps produced by the DBSN. **(a)** The input original image. **(b)** Four feature maps produced by the layer with 32 channels of each branch. **(bi)** Feature maps of the encoder branch (*In\_encoder*); **(bii)** Feature maps of the upper decoder branch (*In\_upper*); **(biii)** Feature maps of the lower decoder branch (*In\_lower*); **(biv)** Feature maps of the attention gate (*In\_ag*); **(c)** The corresponding ground truth (GT).



**Supplemental Fig. 2**. Four different architectures for the ablation study. **(a)** Our Double Branch Segmentation Network (DBSN) model contains the lower branch (LB), attention gate (AG) and deformable convolution (DC). **(b)** DBSN without AG. **(c)** DBSN without LB. **(d)** DBSN without DC. Note: the purple line indicates Up-sampling + 3$×$3 DC + Group-normalization + ReLU, whereas the green line indicates Up-sampling + 3$×$3 convolution + Group-normalization + ReLU. The green and blue rectangular blocks respectively represent feature maps of the upper branch and LB.