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Algorithm 1: Self-Registration(E,T)
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Input: E: the subject's epi sequence.
  T: the subject's anatomical scan.
  Result: E_T: the subject's epi sequence in anatomical space.
  // FLIRT using local optimisation
1 T_b = bet(T, -f0.3 - R - B - S)
\mathbf{z} xfm1 = flirt(E, T_b, sch = simple 3d.sch)
E_{local} = applyxfm(E, T_b, xfm1)
4 E_{bbr} = epireg(E_{local}, T, T_b)
5 if regscore(A_{bbr}, T_b) > 0.8 then
      // Our nonlinear method has not completely failed
      return resample(E_{bbr}, T_b)
6
7 else
      // Use the more robust method, which was the input to the previous
     return resample(E_{local}, T_b)
9 end
```

```
Algorithm 2: Template-Registration(E,T,M)
   Input: E: the subject's epi sequence in anatomical space.
   T: the subject's anatomical scan.
   M: the template; contains a brain and mask attribute, and M itself refers to the skull-on
  image.
  Result: E_T: the subject's epi sequence in anatomical space.
   // FLIRT using local optimisation
 1 T_b = bet(T, -f0.3 - R - B - S)
\mathbf{z} \ xfm1 = flirt(E, T_b)
 \mathbf{3} \ warp1 = fnirt(T, M, guess = xfm1, mask = M.mask)
 4 E_{nonlin} = applywarp(E, M, guess = warp1)
5 if regscore(E_{nonlin}, M_b) > 0.8 then
      // Our nonlinear method has not completely failed
 6
      T = resample(T, M_b)
      return resample(E_{nonlin}, M_b)
7
8 else
      // Use the more robust method, so compute here
      E_{lin} = applyxfm(T, M, guess = xfm1)
9
      T = applyxfm(T, M, guess = xfm1)
10
      resample(T, M_b)
11
      return resample(E_{lin}, M_b)
13 end
```

## **Algorithm 3:** Reg-Score(A, $B_M$ )

**Input:** A: the aligned sequence. note that this image can be skullstripped or not since we skullstrip internally.

 $B_M$ : the brain that was aligned to. note that we assume this input has already been skull stripped and ias a mask.

**Result:** r: the registration score between A and B.

```
// Generate an approximate mask for our data 1 A_M=bet(A,-f0.3-m) 2 return \frac{A_M\cap B_M}{A_M\cup B_M}
```