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#include <cstddef>
#include <stdlib.h>
#include "DGM.h"

using namespace DirectGraphicalModels;

class Foo {

public:

    Foo(byte n_states, Size img_size):
        dense_graph(n_states) ,
        inferer(dense_graph),
        ext_graph(dense_graph)
    {

        /*Arguments:
           n_states: the number of depth states.
           img_size: size of image as (cols,rows).
        */

        // build a 2D graph with image size and default potentials
        ext_graph.buildGraph(img_size);
    }

    ~Foo(void) = default;

    void add_unary_pot(Mat unary_pot);

    void set_smooth_kernel(Vec2f sigma_smooth, float w_smooth);

    void add_appearance_kernel(Mat feature, Vec2f sigma_xy, float sigma_feat, float
w_kernel);

    CGraphDense& getGraph() { return dense_graph; }

    CIferDense& getInfer() { return inferer; }

    CGraphDenseExt& getGraphExt() { return ext_graph; }

private:
    CGraphDense dense_graph;
    CIferDense inferer;
    CGraphDenseExt ext_graph;
};

#include "depth_regression.h"

void Foo::add_unary_pot(Mat unary_pot) {

```

```

/* Arguments:
   unary_pot: unary potential matrix with the shape of (n_nodes, n_states).
*/
// adds the graph nodes with potentials.
getGraph().addNodes(unary_pot);

return;

}

void Foo::set_smooth_kernel(Vec2f sigma_xy, float w_kernel){
    /*Arguments:
       sigma_xy: x-y std deviations of the 2D smoothness kernel.
       w_kernel: weighting parameter of the smoothness term.
    */
    // add smoothness potential
    getGraphExt().addGaussianEdgeModel(sigma_xy, w_kernel);

    return;
}

void Foo::add_appearance_kernel(Mat feature, Vec2f sigma_xy, float sigma_feat, float w_kernel){
    /*Arguments:
       feature: a multi-channel feature matrix with shape of (cols,rows).
       sigma_xy: x-y std deviations of the 2D smoothness kernel,
       sigma_feat: std deviation of the feature.
       w_kernel: weighting parameter of the smoothness term.
    */
    // add a Bilateral pairwise potential
    getGraphExt().addBilateralEdgeModel(feature, sigma_xy, sigma_feat, w_kernel);

    return;
}

```

Unit Test:

```

#include<iostream>
#include<string>
#include "depth_regression.h"

void show(std::string name, Mat img)
{
    namedWindow(name, WINDOW_NORMAL);
    imshow(name, img);
}

```

```

    return;
}

int main(void){

    byte n_states = 1;

    const int cols = 9;

    const int rows = 9;

    const Size img_size = Size (cols,rows);

    Mat gt_seg(rows, cols, CV_32FC1, Scalar(0));

    // draw a 3x3 rectangle
    for( int i = 3; i<6; i++){

        float * row_i = gt_seg.ptr<float>(i);

        for(int j=3; j<6; j++){

            row_i[j] = 1;

        }

    }

    Mat gt_img;
    gt_seg.convertTo(gt_img, CV_8U, 255, 0);

    show("gt_img", gt_img);

    Foo fcrf {n_states, img_size};

    Mat unary_pot;
    // blur gt_seg to get foreground predictions
    GaussianBlur(gt_seg, unary_pot, Size(3, 3), 1.0, 1.0);

    show("unary_pot", unary_pot);

    unary_pot = unary_pot.reshape(1, cols*rows);

    fcrf.add_unary_pot(unary_pot);

    fcrf.set_smooth_kernel(Vec2f(3,3), 1);

    Mat features;
    // obtain features
    Canny(gt_img, features, 0, 1, 7);
}

```

```
show("features", features);

fcrf.add_appearance_kernel(features, Vec2f(3,3), 5, 2);

// estimates the marginal potentials for each graph node
vec_byte_t optimalDecoding = depth_fcrf.getInfer().decode(10);

waitKey(0);

return 0;

}
```