# A GPU simulation of skyrmion

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#### I. LLG

The Landau-Lifshitz-Gilbert-Equation (LLG) can be written as (Eq. (13) of Ref. [1] and Eq. (7) of Ref. [2], NOTE that the sign of the last term is '+' in Eq. (7) of Ref. [2])

$$\dot{\mathbf{n}} = \gamma \mathbf{B}_{\text{eff}} \times \mathbf{n} - \frac{\alpha \gamma}{|n|} \mathbf{n} \times \dot{\mathbf{n}} - \frac{\hbar \gamma}{2e} \left( \mathbf{j} \cdot \nabla \right) \mathbf{n}$$
(1)

where **n** is the magnetic momentum,  $\gamma(\gamma > 0)$  is the gyromagnetic ratio,  $\alpha$  represents Gilbert damping,  $\mathbf{B}_{\text{eff}}$  is the effective field arising from the spin Hamiltonian, and can be written as  $(H_S \text{ is from Ref. [2]}, \text{ before Eq. (1)}, \text{ also Eq. (4) in Ref. [3]})$ 

$$\mathbf{B}_{\text{eff}} \equiv \frac{\delta H_S}{\delta \mathbf{n}}$$

$$H_S = \int d^D x \frac{J}{2a} (\nabla \mathbf{n})^2 + \frac{D}{a^2} \mathbf{n} \cdot (\nabla \times \mathbf{n}) - \frac{\mu}{a^3} \mathbf{B} \cdot \mathbf{n}$$
(2)

# II. LLG IN 2D LATTICE

In the following, for simplicity, we use  $\delta = a(\mathbf{e}_x, \mathbf{e}_y)$ ,  $\delta_x = a\mathbf{e}_x$ ,  $\delta_y = a\mathbf{e}_y$ , where a is the distance between two lattice.

## A. Spin torque

Written in lattice, so that

$$(\mathbf{j} \cdot \nabla) \mathbf{n} = \sum_{i} j_{i} \partial_{i} n_{x} \mathbf{e}_{x} + \sum_{i} j_{i} \partial_{i} n_{y} \mathbf{e}_{y} + \sum_{i} j_{i} \partial_{i} n_{z} \mathbf{e}_{z}$$
(3)

In 2D, it is

$$(\mathbf{j} \cdot \nabla) \mathbf{n} = \sum_{i=x,y} j_i (\partial_i n_x \mathbf{e}_x + \partial_i n_y \mathbf{e}_y + \partial_i n_z \mathbf{e}_z)$$

$$= \frac{1}{a} \sum_{i=x,y} j_i (\frac{n_x (\mathbf{r} + \delta_i) - n_x (\mathbf{r} - \delta_i)}{2} \mathbf{e}_x + \frac{n_y (\mathbf{r} + \delta_i) - n_y (\mathbf{r} - \delta_i)}{2} \mathbf{e}_y + \frac{n_z (\mathbf{r} + \delta_i) - n_z (\mathbf{r} - \delta_i)}{2} \mathbf{e}_z)$$

$$= \frac{1}{a} \sum_{i=x,y} j_i \frac{\mathbf{n}(\mathbf{r} + \delta_i) - \mathbf{n}(\mathbf{r} - \delta_i)}{2}$$

$$(4)$$

Sometimes, it is also written as (see last term in Eq. (8) in Ref. [4], it also says this is the discrete version of the continuous term  $(\mathbf{j} \cdot \nabla)\mathbf{n}$  before Eq. (10))

$$(\mathbf{j} \cdot \nabla) \mathbf{n} = \frac{1}{a} \sum_{i=x,y} j_i \mathbf{n}(\mathbf{r}) \times \left( \frac{\mathbf{n}(\mathbf{r} + \delta_i) - \mathbf{n}(\mathbf{r} - \delta_i)}{2} \times \mathbf{n}(\mathbf{r}) \right)$$
(5)

that is because, for a unit vector, one have

$$\mathbf{n} \cdot \partial_i \mathbf{n} = 0 \tag{6}$$

the discrete version is

$$\mathbf{n}(\mathbf{r}) \cdot \frac{\mathbf{n}(\mathbf{r} + \delta_{\mathbf{i}}) - \mathbf{n}(\mathbf{r} - \delta_{\mathbf{i}})}{2\mathbf{a}} = \mathbf{0}$$
(7)

so that

$$\mathbf{n}(\mathbf{r}) \times \left( \frac{\mathbf{n}(\mathbf{r} + \delta_i) - \mathbf{n}(\mathbf{r} - \delta_i)}{2} \times \mathbf{n}(\mathbf{r}) \right)$$

$$= \frac{\mathbf{n}(\mathbf{r} + \delta_i) - \mathbf{n}(\mathbf{r} - \delta_i)}{2} (\mathbf{n}(\mathbf{r}) \cdot \mathbf{n}(\mathbf{r})) - \mathbf{n}(\mathbf{r}) \left( \mathbf{n}(\mathbf{r}) \cdot \frac{\mathbf{n}(\mathbf{r} + \delta_i) - \mathbf{n}(\mathbf{r} - \delta_i)}{2} \right)$$

$$= \frac{\mathbf{n}(\mathbf{r} + \delta_i) - \mathbf{n}(\mathbf{r} - \delta_i)}{2}$$
(8)

to be consist with the references, we use

$$(\mathbf{j} \cdot \nabla) \mathbf{n} = \frac{1}{a} \sum_{i=x,y} j_i \mathbf{n}(\mathbf{r}) \times \left( \frac{\mathbf{n}(\mathbf{r} + \delta_i) - \mathbf{n}(\mathbf{r} - \delta_i)}{2} \times \mathbf{n}(\mathbf{r}) \right)$$
(9)

In simulation, to simplify, we always take the direction of  $\mathbf{j}$  as  $\mathbf{x}$ -axis, so we only have

$$\frac{1}{a}j_x\mathbf{n}(\mathbf{r})\times\left(\frac{\mathbf{n}(\mathbf{r}+\delta_x)-\mathbf{n}(\mathbf{r}-\delta_x)}{2}\times\mathbf{n}(\mathbf{r})\right)$$
(10)

#### B. J term

 $(\nabla \mathbf{n})^2$  is confusing, in 2D, it is in fact  $\sum_{i=x,y} (\partial_i \mathbf{n}) \cdot (\partial_i \mathbf{n})$  Consider in 1-dimension, we have

$$\mathbf{n}(x) \cdot \partial_{i} \mathbf{n}(x) = 0$$

$$0 = \sum_{i=x,y} \partial_{i} (\mathbf{n}(x) \cdot \partial_{i} \mathbf{n}(x)) = \sum_{i=x,y} (\partial_{i} \mathbf{n}) \cdot (\partial_{i} \mathbf{n}) + \mathbf{n} \cdot (\sum_{i=x,y} (\partial_{i})^{2}) \mathbf{n}$$
(11)

In lattice, the derivate can be written as

$$\frac{d^2}{dx^2}\mathbf{n}(\mathbf{r}) = \frac{1}{a}\left(\mathbf{n}'(\mathbf{r} + \frac{a}{2}\mathbf{e}_x) - \mathbf{n}'(\mathbf{r} - \frac{a}{2}\mathbf{e}_x)\right)$$
(12)

with

$$\mathbf{n}'(\mathbf{r} + \frac{a}{2}\mathbf{e}_x) = \frac{1}{a}(\mathbf{n}(\mathbf{r} + a\mathbf{e}_x) - \mathbf{n}(\mathbf{r}))$$

$$\mathbf{n}'(\mathbf{r} - \frac{a}{2}\mathbf{e}_x) = \frac{1}{a}(\mathbf{n}(\mathbf{r}) - \mathbf{n}(\mathbf{r} - a\mathbf{e}_x))$$
(13)

so that

$$\left(\sum_{i=x,y} (\partial_i)^2\right) \mathbf{n} = \frac{1}{a^2} \sum_{i=x,y,-x,-y} \mathbf{n}(\mathbf{r} + \delta_i) - 4\mathbf{n}(\mathbf{r})$$
(14)

and

$$\mathbf{n} \cdot \left(\sum_{i=x,y} (\partial_i)^2\right) \mathbf{n} = \frac{1}{a^2} \sum_{\langle i \rangle} \mathbf{n} \cdot \mathbf{n}_i - 4 \tag{15}$$

Throw away the constant term

$$\int d^D x \frac{J}{2a} (\nabla \mathbf{n})^2 = -\frac{J}{2a^3} \sum_{\langle i,j \rangle} \mathbf{n}_i \cdot \mathbf{n}_j$$
(16)

where j are all neighbours of  $i, j = i + \delta_x, i - \delta_x, i + \delta_y, i - \delta_y$ . When J is constant, it can also been written as

$$\int d^D x \frac{J}{2a} (\nabla \mathbf{n})^2 = -\frac{J}{a^3} \sum_{\mathbf{r}} \mathbf{n}(\mathbf{r}) \cdot (\mathbf{n}(\mathbf{r} + \delta_x) + \mathbf{n}(\mathbf{r} + \delta_y))$$
(17)

C. D term

In 2D, we find

$$\mathbf{n} \cdot (\nabla \times \mathbf{n}) = n_x \partial_y n_z - n_y \partial_x n_z + n_z \partial_x n_y - n_z \partial_y n_x$$

$$= \frac{1}{2a} \left[ (n_x(\mathbf{r})(n_z(\mathbf{r} + \delta_y) - n_z(\mathbf{r} - \delta_y)) - n_z(\mathbf{r})(n_x(\mathbf{r} + \delta_y) - n_x(\mathbf{r} - \delta_y))) \right]$$

$$+ (n_z(\mathbf{r})(n_y(\mathbf{r} + \delta_x) - n_y(\mathbf{r} - \delta_x)) - n_y(\mathbf{r})(n_z(\mathbf{r} + \delta_x) - n_z(\mathbf{r} - \delta_x))) \right]$$

$$= \frac{1}{2a} \left[ (n_x(\mathbf{r})n_z(\mathbf{r} + \delta_y) - n_z(\mathbf{r})n_x(\mathbf{r} + \delta_y)) - (n_x(\mathbf{r})n_z(\mathbf{r} - \delta_y) - n_z(\mathbf{r})n_x(\mathbf{r} - \delta_y)) \right]$$

$$+ (n_z(\mathbf{r})n_y(\mathbf{r} + \delta_x) - n_y(\mathbf{r})n_z(\mathbf{r} + \delta_x)) - (n_z(\mathbf{r})n_y(\mathbf{r} - \delta_x) - n_y(\mathbf{r})n_z(\mathbf{r} - \delta_x)) \right]$$

$$= -\frac{1}{2a} \left[ \mathbf{n}(\mathbf{r}) \cdot (\mathbf{n}(\mathbf{r} + \delta_y) \times \mathbf{e}_y) - \mathbf{n}(\mathbf{r}) \cdot (\mathbf{n}(\mathbf{r} - \delta_y) \times \mathbf{e}_y) \right]$$

$$= -\frac{1}{2a^2} \left[ \mathbf{n}(\mathbf{r}) \times \mathbf{n}(\mathbf{r} + \delta_y) \cdot \delta_y - \mathbf{n}(\mathbf{r}) \times \mathbf{n}(\mathbf{r} - \delta_y) \cdot \delta_y \right]$$

$$+ \mathbf{n}(\mathbf{r}) \times \mathbf{n}(\mathbf{r} + \delta_x) \cdot \delta_x - \mathbf{n}(\mathbf{r}) \times \mathbf{n}(\mathbf{r} - \delta_x) \cdot \delta_x \right]$$

so, one have

$$\int d^{D}x \frac{D}{a^{2}} \mathbf{n} \cdot (\nabla \times \mathbf{n}) = -\frac{D}{a^{4}} \sum_{\mathbf{r}} (\mathbf{n}(\mathbf{r}) \times \mathbf{n}(\mathbf{r} + \delta_{x}) \cdot \delta_{x} + \mathbf{n}(\mathbf{r}) \times \mathbf{n}(\mathbf{r} + \delta_{y}) \cdot \delta_{y})$$
(19)

It is also

$$\int d^D x \frac{D}{a^2} \mathbf{n} \cdot (\nabla \times \mathbf{n}) = -\frac{D}{a^4} \sum_{\mathbf{r}} (\mathbf{n}(\mathbf{r} + \delta_x) \times \delta_x + \mathbf{n}(\mathbf{r} + \delta_y) \times \delta_y) \cdot \mathbf{n}(\mathbf{r})$$
(20)

## D. Effective magnetic field

Using the results, we can write the discrete version of  $H_s$  as

$$H_S = \sum_{\mathbf{r}} \left[ -\frac{J}{a^3} (\mathbf{n}(\mathbf{r} + \delta_x) + \mathbf{n}(\mathbf{r} + \delta_y)) - \frac{D}{a^3} (\mathbf{n}(\mathbf{r} + \delta_x) \times \mathbf{e}_x + \mathbf{n}(\mathbf{r} + \delta_y) \times \mathbf{e}_y) - \frac{\mu}{a^3} \mathbf{B} \right] \cdot \mathbf{n}(\mathbf{r})$$
(21)

which leads to

$$\mathbf{B}_{\text{eff}}(\mathbf{r}) = \frac{\delta H_S}{\delta \mathbf{n}} = -\frac{1}{a^3} \sum_{i=x,y} \left[ J(\mathbf{r}) \mathbf{n} (\mathbf{r} + \delta_i) + J(\mathbf{r} - \delta_i) \mathbf{n} (\mathbf{r} - \delta_i) \right] - \frac{1}{a^3} \sum_{i=x,y} \left[ D(\mathbf{r}) \mathbf{n} (\mathbf{r} + \delta_i) \times \mathbf{e}_i - D(\mathbf{r} - \delta_i) \mathbf{n} (\mathbf{r} - \delta_i) \times \mathbf{e}_i \right] - \frac{\mu}{a^3} \mathbf{B}(\mathbf{r})$$
(22)

#### E. anisotropy

In some material, the effective Hamiltonian can be written with anisotropy terms as (ignore a, this is as same as Eq. (10) in Ref. spintransfer)

$$H_{S} = \sum_{\mathbf{r}} \left[ -\frac{J}{a^{3}} (\mathbf{n}(\mathbf{r} + \delta_{x}) + \mathbf{n}(\mathbf{r} + \delta_{y})) - \frac{D}{a^{3}} (\mathbf{n}(\mathbf{r} + \delta_{x}) \times \mathbf{e}_{x} + \mathbf{n}(\mathbf{r} + \delta_{y}) \times \mathbf{e}_{y}) - \frac{\mu}{a^{3}} \mathbf{B} \right]$$

$$-\mathbf{h} \cdot \mathbf{n}(\mathbf{r}) - K \left( \mathbf{e}_{z} \cdot \mathbf{n}(\mathbf{r}) \right)^{2}$$
(23)

The contribution of  $\mathbf{h}$  can be just put into the applied magnetic field  $\mathbf{B}$ , we also include the contribution of K.

#### F. dimensionless LLG

Using dimensionless parameters, the LLG can be written as

$$\dot{\mathbf{n}} = \mathbf{B}_{\text{eff}} \times \mathbf{n} - \alpha \mathbf{n} \times \dot{\mathbf{n}} - \sum_{i=x,y} j_{i} \mathbf{n}(\mathbf{r}) \times \left( \frac{\mathbf{n}(\mathbf{r} + \delta_{i}) - \mathbf{n}(\mathbf{r} - \delta_{i})}{2} \times \mathbf{n}(\mathbf{r}) \right)$$

$$\mathbf{B}_{\text{eff}}(\mathbf{r}) = \frac{\delta H_{S}}{\delta \mathbf{n}} = -\sum_{i=x,y} \left[ J(\mathbf{r}) \mathbf{n}(\mathbf{r} + \delta_{i}) + J(\mathbf{r} - \delta_{i}) \mathbf{n}(\mathbf{r} - \delta_{i}) \right]$$

$$-\sum_{i=x,y} \left[ D(\mathbf{r}) \mathbf{n}(\mathbf{r} + \delta_{i}) \times \mathbf{e}_{i} - D(\mathbf{r} - \delta_{i}) \mathbf{n}(\mathbf{r} - \delta_{i}) \times \mathbf{e}_{i} \right] - \mathbf{B}(\mathbf{r}) - 2K \left( \mathbf{e}_{z} \cdot \mathbf{n}(\mathbf{r}) \right) \mathbf{e}_{z}$$
(24)

Ignoring the anisotropy term, this is as same as Eq. (9) in Ref. [4].

## III. EVALUATION

Let 
$$\mathbf{N} = \mathbf{B}_{\text{eff}} \times \mathbf{n} - \sum_{i=x,y} j_i \mathbf{n}(\mathbf{r}) \times \left( \frac{\mathbf{n}(\mathbf{r} + \delta_i) - \mathbf{n}(\mathbf{r} - \delta_i)}{2} \times \mathbf{n}(\mathbf{r}) \right)$$
, we have 
$$\dot{\mathbf{n}} = \mathbf{N} - \alpha \mathbf{n} \times \dot{\mathbf{n}}$$
(25)

This is in fact a combine of 3 equations which can be written as

$$\frac{dn_x}{dt} = \frac{1}{1+\alpha^2} \left( \alpha^2 n_x^2 N_x + \alpha (\alpha n_x (n_y N_y + n_z N_z) - n_y N_z + N_y n_z) + N_x \right), \dots$$
 (26)

Assuming  $\alpha \ll 1$ , it can be written as

$$\frac{d\mathbf{n}}{dt} = \mathbf{N} + \alpha \mathbf{N} \times \mathbf{n} + \mathcal{O}(\alpha^2) \tag{27}$$

One can use this to evaluate

$$\mathbf{n}(t + \Delta t) = \mathbf{n}(t) + \Delta t \left( \mathbf{N}(t) + \alpha \mathbf{N}(t) \times \mathbf{n}(t) \right)$$
(28)

## IV. SIMULATION

The simulation is running on GPU using the compute shader in Unity3D. The implementation of LLG is the LLG\_c.compute, the content is

```
// Each #kernel tells which function to compile; you can have many kernels
   #pragma kernel CSMain
   // Create a RenderTexture with enableRandomWrite flag and set it
   // with cs.SetTexture
   //RWStructuredBuffer<float3> magneticMomentum;
   RWTexture2D<float4> magneticMomentum;
   Texture2D<float4> boundaryCondition;
   Texture2D<float4> exchangeStrength;
10
   uint2 size;
   float K;
   float D:
   float B;
   float jx;
   float alpha;
   float timestep;
    [numthreads(8, 8, 1)]
   void CSMain (uint3 id : SV_DispatchThreadID)
   {
21
```

```
float4 zero4 = float4(0.5f, 0.5f, 0.5f, 1.0f);
22
       float3 s = magneticMomentum[id.xy].xyz;
23
       float3 sleft = id.x > 1 ? magneticMomentum[id.xy - uint2(1, 0)].xyz : zero4.xyz;
24
       float3 sright = id.x < (size.x - 1) ? magneticMomentum[id.xy + uint2(1, 0)].xyz : zero4.xyz;</pre>
25
       float3 sup = id.y > 1 ? magneticMomentum[id.xy - uint2(0, 1)].xyz : zero4.xyz;
       float3 sdown = id.y < (size.y - 1) ? magneticMomentum[id.xy + uint2(0, 1)].xyz : zero4.xyz;</pre>
27
       s = 2.0f * (s - 0.5f);
28
       sleft = 2.0f * (sleft - 0.5f);
       sright = 2.0f * (sright - 0.5f);
30
31
       \sup = 2.0f * (\sup - 0.5f);
       sdown = 2.0f * (sdown - 0.5f);
32
33
       float edge = boundaryCondition[id.xy].r > 0.5f ? 1.0f : 0.0f;
34
35
       float j_s = exchangeStrength[id.xy].x;
36
37
        float j_left = id.x > 1 ? exchangeStrength[id.xy - uint2(1, 0)].x : 0.0f;
       float j_up = id.y > 1 ? exchangeStrength[id.xy - uint2(0, 1)].x : 0.0f;
38
39
       float3 vright = float3(1.0, 0.0, 0.0);
40
       float3 vdown = float3(0.0, 1.0, 0.0);
41
42
       float3 beff = (j_left * sleft + j_s * sright + j_up * sup + j_s * sdown)
43
           + D * (cross(sright, vright) - cross(sleft, vright) + cross(sdown, vdown) - cross(sup, vdown))
44
           + float3(0.0f, 0.0f, B) + 2 K * float3(0.0f, 0.0f, s.z);
46
47
       float3 stt = -jx * cross(s, cross((sright - sleft) * 0.5f, s));
48
       float3 newS = cross(s, beff) + stt;
49
       newS = newS - alpha * cross(s, newS);
50
51
52
       float3 retColor = normalize(s + timestep * newS) * 0.5f + 0.5f;
53
       magneticMomentum[id.xy] = float4(retColor.r, retColor.g, retColor.b, 1.0f) * edge + (1.0f - edge) * zero4;
   }
54
```

The magnetic momentum is a 512 × 512 64-bit ARGB texture, only R, G, B channel used.  $\mathbf{n} = (2 \times r - 1, 2 \times g - 1, 2 \times b - 1)$ .

The boundary condition is a  $512 \times 512$  alpha 8-bit texture, only R channel used, when R < 0.5, it is a defect.

The exchange strength is a 32-bit RFloat texture, generated from a Lua script. For example, a constant exchange strength can be generated from a lua file as

```
1  -- Exchange Strength is constant
2  function GetJValueByLatticeIndex(x, y)
3    return 2.0
4  end
5
6  -- Need to register the function
7  return {
8   GetJValueByLatticeIndex = GetJValueByLatticeIndex,
9 }
```

While a pin with  $J = 1 + \exp(-0.001\rho^2)$  at lattice index (255, 255) can be written as

```
1
    -- Exchange Strength is pin
    function GetJValueByLatticeIndex(x, y)
       local j0 = 1
3
       local j1 = 1
4
5
       local j2 = 0.001
       local rho = (x - 255) * (x - 255) + (y - 255) * (y - 255)
6
       return j0 + j1 * math.exp(-1.0 * j2 * rho)
8
   end
9
     -- Need to register the function
11
12
   return {
       GetJValueByLatticeIndex = GetJValueByLatticeIndex,
13
14
```

Manual.pdf is a document introduce how to use the pre-built software.

- [1] Gen Tatara, Hiroshi Kohno, Junya Shibata, Phys. Rep. 468, 213-301 (2008), 10.1016/j.physrep.2008.07.003, arXiv:0807.2894.
- [2] Jiadong Zang, Maxim Mostovoy, Jung Hoon Han, and Naoto Nagaosa, 10.1103/PhysRevLett.107.136804.
- [3] Hong Chul Choi, Shi-Zeng Lin, Jian-Xin Zhu, Phys. Rev. B 93, 115112 (2016), 10.1103/PhysRevB.93.115112, arXiv:1601.00933.
- $[4] \ \ Ye-Hua\ Liu,\ You-Quan\ Li,\ J.\ Phys.:\ Condens.\ Matter\ 25\ 076005,\ 10.1088/0953-8984/25/7/076005,\ arXiv:1206.5661.$
- [5] Junichi Iwasaki, Wataru Koshibae, and Naoto Nagaosa, Nano. Lett. 2014, 14, 4432-4437, 10.1021/nl501379k.