

Capsule Network with Routing Mechanism

Part 2: Matrix Capsule & EM Routing

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Agenda

1. Recap & Matrix Capsule Network

- (Vector) Capsules
- Dynamic Routing by Agreement
- (Recap: Capsule blueprint & Routing by Agreement
- Matrix) Capsules Blueprint

2. Routing Mechanism and Unsupervised Clustering

- Dynamic Routing & k-Mean
- GMM & EM Routing & Gaussian Mixture Model

3. Experiments

- smallNORB classification task
- Adversarial examples

(Vector)**Capsules** blueprint

- "A capsule is a group of neurons whose output represents different properties of the same entity."
- General ideas differ from [Sabour et al. 2017]:
 - \circ Vector \rightarrow Matrix
 - $\circ \quad \text{Activity Vector} \rightarrow \text{Pose Matrix} + \text{Activity Probability}$





[Hinton, G. E., Sabour, S., Frosst, N. (2018). Matrix Capsules with EM Routing.]



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(Matrix) Capsule Network Blueprint



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Routing by EM Clustering (GMM)



Architecture: Matrix Capsule



[Hinton, G. E., Sabour, S., Frosst, N. (2018). Matrix Capsules with EM Routing. ICLR 2018]

Experiments: smallNORB

| Routing iterations | Pose structure | Loss | Coordinate Addition | Test error rate |
|-------------------------------------|----------------|---------------------|---------------------|-----------------|
| 1 | Matrix | Spread | Yes | 9.7% |
| 2 | Matrix | Spread | Yes | 2.2% |
| 3 | Matrix | Spread | Yes | 1.8% |
| 5 | Matrix | Spread | Yes | 3.9% |
| 3 | Vector | Spread | Yes | 2.9% |
| 3 | Matrix | Spread | No | 2.6% |
| 3 | Vector | Spread | No | 3.2% |
| 3 | Matrix | Margin ¹ | Yes | 3.2% |
| 3 | Matrix | CrossEnt | Yes | 5.8% |
| Descline CNN with 4 2M person store | | | | 5.007 |

Baseline CNN with 4.2M parameters 5.2% CNN of Ciresan et al. (2011) with extra input images & deformations 2.56%

1.4%

Our Best model (third row), with multiple crops during testing

Open Source Implementation:

- CNN baseline (4.2M): 88.7% (best)/94.8% (paper)
- Matrix Cap with EM routing (310K, 2 iteration): 91.8%(best)/98.6%(paper)
- https://github.com/www0wwwjs1/Matrix-Capsules-EM-Tensorflow

[Hinton, G. E., Sabour, S., Frosst, N. (2018). Matrix Capsules with EM Routing. ICLR 2018]



https://cs.nyu.edu/~ylclab/data/norb-v1.0-small/

Experiments: smallNORB



[Hinton, G. E., Sabour, S., Frosst, N. (2018). Matrix Capsules with EM Routing. ICLR 2018]

Experiments: Adversarial Robustness



*BIM & FGSM are methods for creating adversarial examples



BIM CNN
BIM CAPS

[Hinton, G. E., Sabour, S., Frosst, N. (2018). Matrix Capsules with EM Routing. ICLR 2018]

Summary of Matrix CapsNet

- Key Points of Matrix Capsule:
 - (Matrix, Activation) \rightarrow (Matrix, Activation)
 - Encapsulate entity or its pattern
 - **Routing** by *agreement* Mechanism

o ...

- Pros:
 - Equivariance
 - Built-in interpretability
 - Adversarial robustness
- Cons:
 - Reproducibility
 - Computational Performance
 - Routing process
 - o ...

References of this Section

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- 2. [Su, J., Vargas, D. V., & Kouichi, S. (2017). One pixel attack for fooling deep neural networks. arXiv:1710.08864.]
- 3. [Hinton, G (2017). What's wrong with convolutional neural nets. <u>https://www.youtube.com/watch?v=</u> <u>Mqt8fs6ZbHk&t=562s</u>]
- 4. [Sabour, S., Frosst, N., & Hinton, G. E. (2017). Dynamic Routing Between Capsules. arXiv:1710.09829.]
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- 6. [Sukhbaatar, S., Weston, J., & Fergus, R. (2015). End-to-end memory networks. In Advances in neural information processing systems (pp. 2440-2448).]
- 7. [Hung-Yi Lee (2017). Capsule. <u>https://www.youtube.com/watch?v=UhGWH3hb3Hk</u>]