# Make Continual Learning Stronger via C-Flat

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## **C-Flat ()** Motivation

Characterizing generalization from loss landscapes to promote CL leads to CL that are:

- A less-than-stellar performance.
- Not exhaustively explored on loss landscape.



**Seeking for flat minima** (Sharpness-aware Minimization) has proven to be a strong training regime for continual learning.

#### Our hypothesis

• Featuring a flatter loss landscape upon sequential arriving tasks can overcome forgetting, thereby ensure CL stability.



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#### Contribution

- We propose Continual Flatness (C-Flat) optimization to Make Continual Learning Stronger .
- We **propose a unified framwork** of C-Flat covering divers CL method categories, and prove that **Flatter is Better** in early all cases.





• Covering all sorts of CL

## C-Flat (♪) Framework



#### Continual Flatness with A Flatter Landscape

- Seeking flat minima that lie in neighborhood
- > Constraining the uniform curvature of the landscape
- A Unified CL Framework using C-Flat
  - Covering all sort of CL, Reg-based, Mem-based, Exp-based
  - Revisiting sharpness-aware minimization

## **C-Flat ()** Make Continual Learning Stronger



#### > Just a line of code

Method	Technology			CIFAR-100			ImageNet-100		Tiny-ImageNet
	Reg.	Mem.	Exp.	B0_Inc5	B0_Inc10	B0_Inc20	B50_Inc10	B50_Inc25	B0_Inc40
Replay [44]		•		58.83	58.87	62.82	63.89	72.18	43.31
w/ C-Flat				59.98 ↑	59.42 ↑	64.71 ↑	63.60 ↓	73.37 ↑	44.95 ↑
iCaRL [43]		•		58.66	59.76	61.13	64.78	77.25	45.70
w/ C-Flat				59.13 ↑	60.40 ↑	62.93 ↑	65.01 ↑	76.22↓	46.08 ↑
WA [64]	٠			63.36	66.76	68.04	73.17	80.81	55.69
w/ C-Flat				65.70 ↑	67.79 ↑	69.16 ↑	73.56 ↑	83.84 ↑	56.06 ↑
PODNet [11]	•	•		48.05	56.01	63.45	83.66	85.95	54.24
w/ C-Flat				49.70 ↑	56.58 ↑	64.37 ↑	84.31 ↑	86.85 ↑	55.13 ↑
DER [57]			•	69.99	71.01	71.40	85.17	87.10	58.63
w/ C-Flat				71.11 ↑	72.08	72.01 ↑	86.64 ↑	87.96 ↑	60.14 ↑
FOSTER [54]	•		•	63.15	66.73	69.70	84.54	87.81	58.80
w/ C-Flat				63.58 ↑	67.34 ↑	70.89 ↑	85.40 ↑	87.81 -	58.88 ↑
MEMO [68]			•	67.42	69.82	69.91	67.28	83.09	58.15
w/ C-Flat				67.56 ↑	69.94 ↑	71.79 ↑	69.34 ↑	83.41 ↑	58.97 ↑
Average Return				+1.04%	+0.66%	+1.34%	+0.62%	+0.90%	+0.81%
Maximum Return				+2.34%	+1.07%	+1.89%	+2.06%	+3.03%	+1.64%

#### Boosting 8 SOTA methods (span all sorts of CL)

**C-Flat (♪)** Visualization



[NeurIPS 2024] "Make Continual Learning Stronger via C-Flat". 😂 🔉: C<sub>b</sub>-a pitch that is one semitone lower than C tune.





## Thank you for listening.