

AdaCompress:

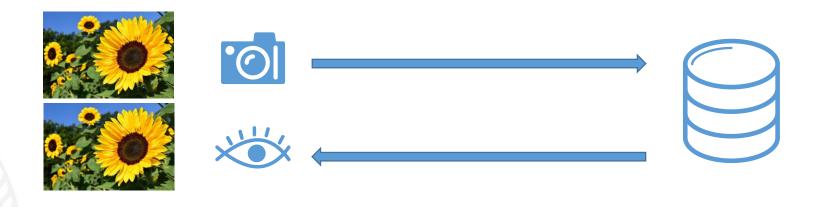
Adaptive Compression for Online Computer Vision Services

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More images are uploaded to DL services rather than human







{c1: {x: 100, ..., class: 101}, c2: {x: 200, ..., class: 203}, c3: {x: 130, ..., class: 303}}

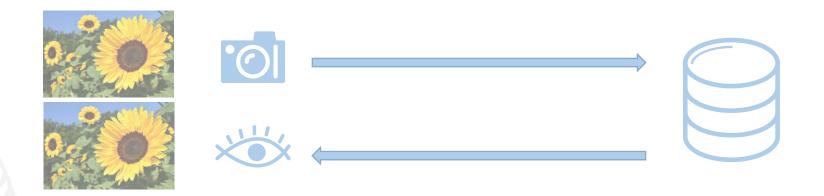








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{c1: {x: 100, ..., class: 101}, c2: {x: 200, ..., class: 203}, c3: {x: 130, ..., class: 303}}



Increasingly Important









Conventional computer vision application framework

- JPEG etc..
- Fixed compression degree for all images
- Same compression strategy for different service providers



• Is the conventional solution efficient enough?





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(1a) Q=75
Face++ prediction = ["donut"]



(1b) Q=55 Face++ prediction = []

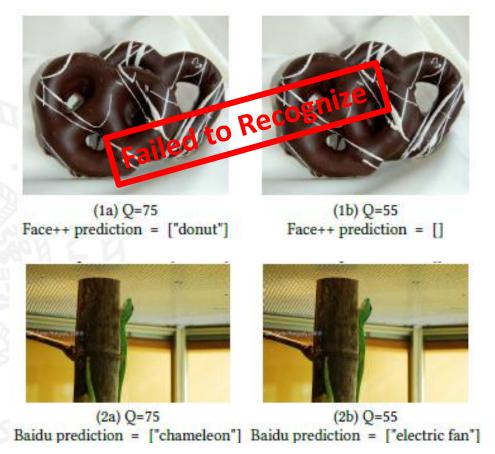


• Is the conventional solution efficient enough?



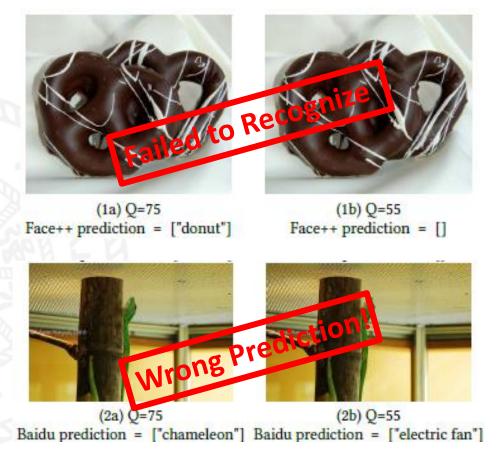


Is the conventional solution efficient enough?



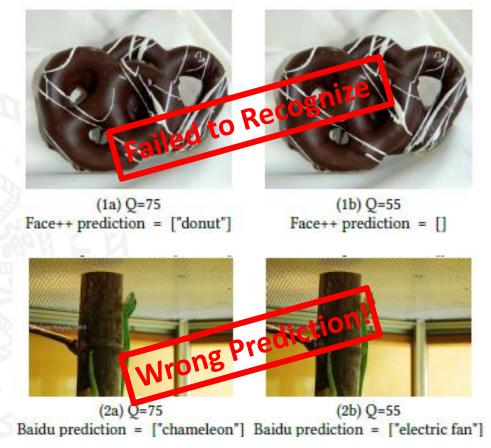


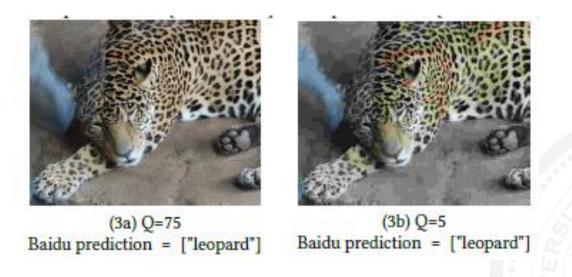
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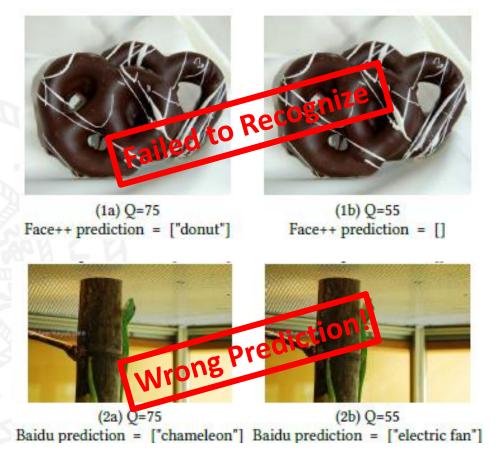
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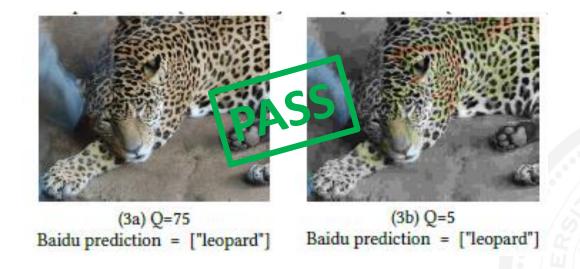






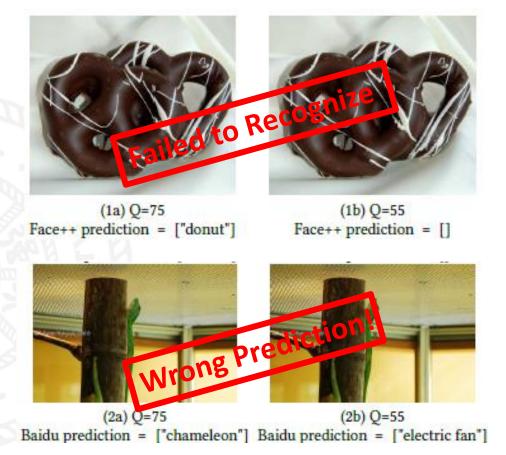
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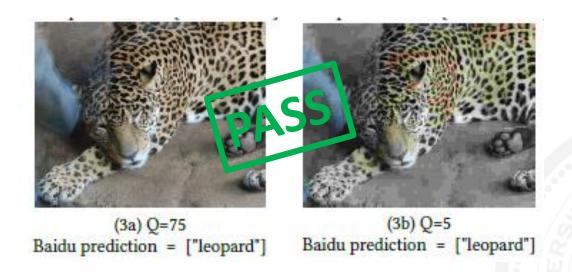






• Is the conventional solution efficient enough?





 In some cases, prediction performance does not related to image quality

Related Works



- Relationship between compression and accuracy
 - Severe compression does not always deteriorate the model inference accuracy (Delac et al.)
 - Four types of quality distortions can affect the performance in deep learning inference (Dodge et al.)

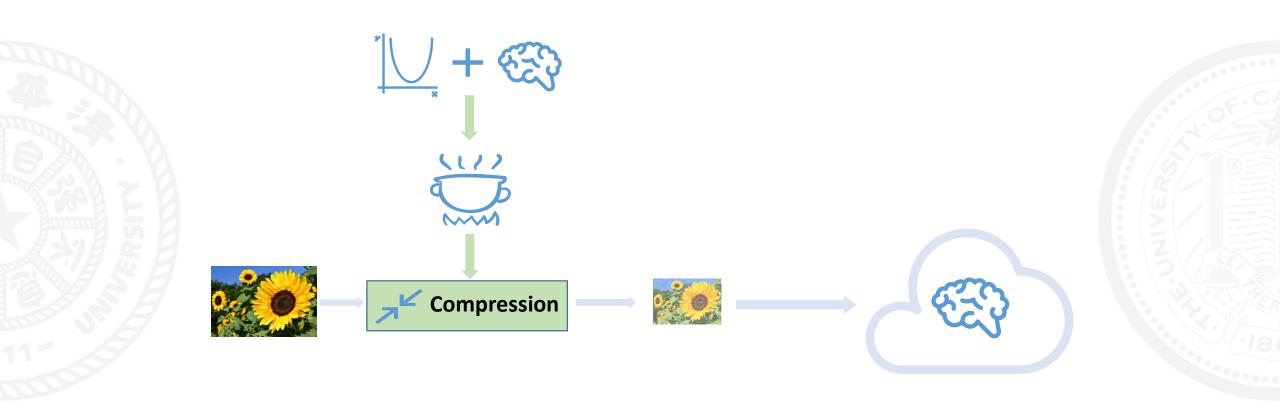
Related Works



- Relationship between compression and accuracy
 - Severe compression does not always deteriorate the model inference accuracy (Delac et al.)
 - Four types of quality distortions can affect the performance in deep learning inference (Dodge et al.)
- Dedicated compression for DNNs
 - Train DNNs from the compressed representations of auto-encoder (Robert et al.)
 - Linear JPEG quantization table learned from the dataset (Liu et al.)
 - DNNs inference from block-wise DCT coefficients in JPEG (Baluja et al.)

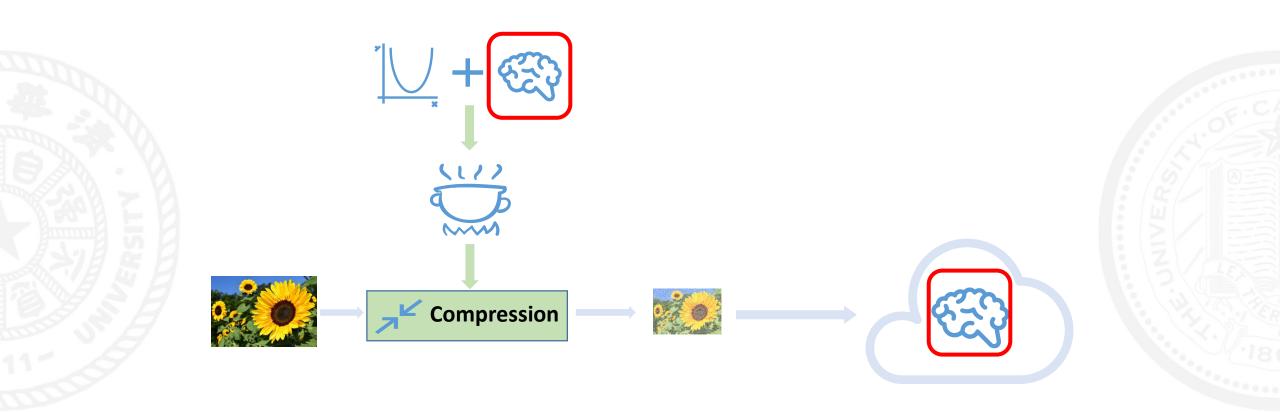


Pre-knowledge of original model



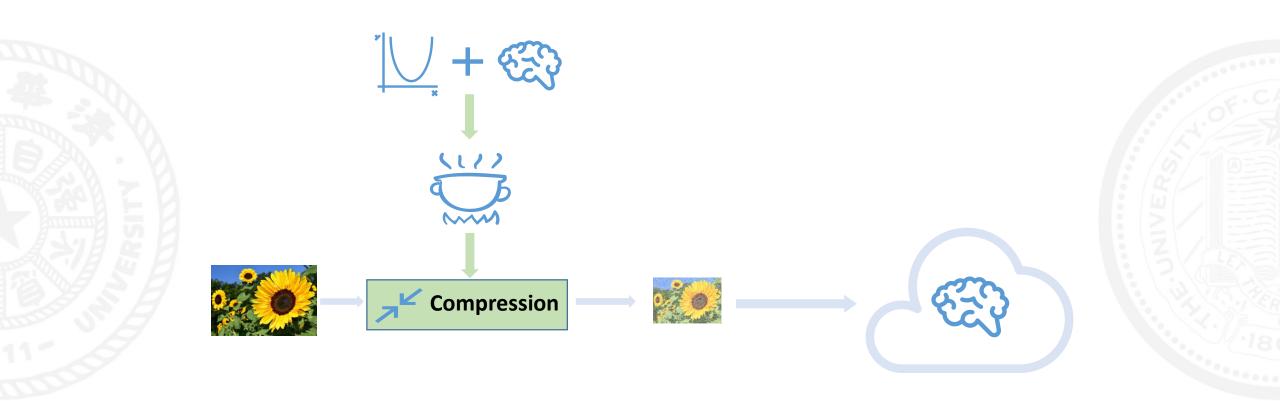


Pre-knowledge of original model



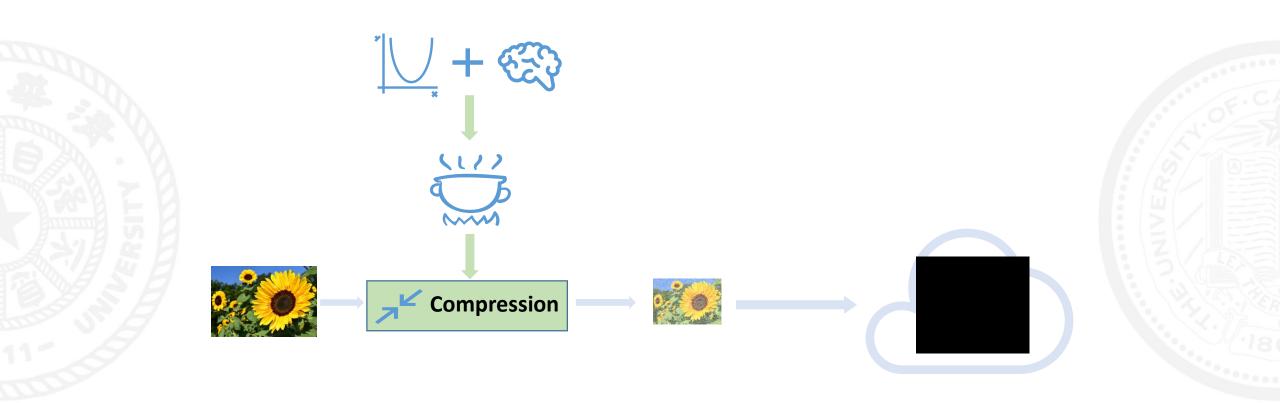


Pre-knowledge of original model



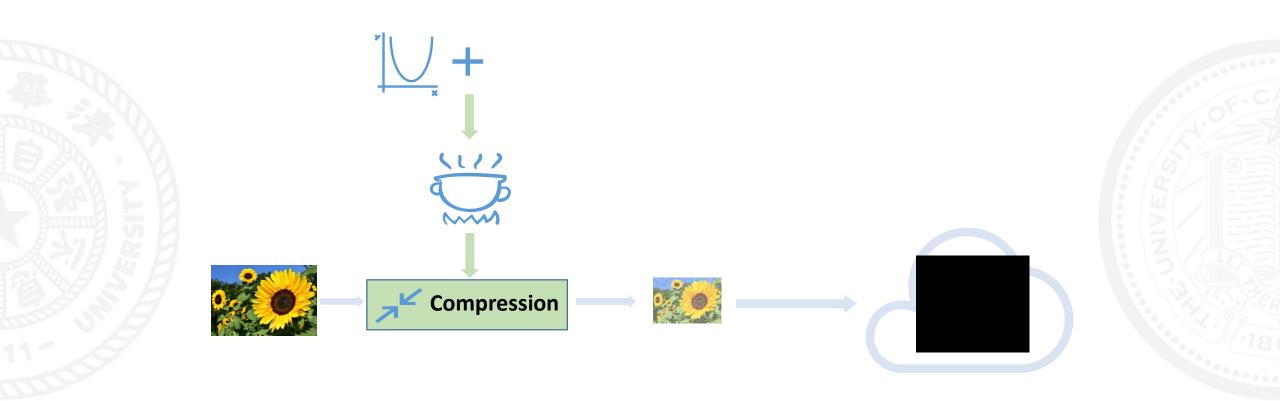


Pre-knowledge of original model

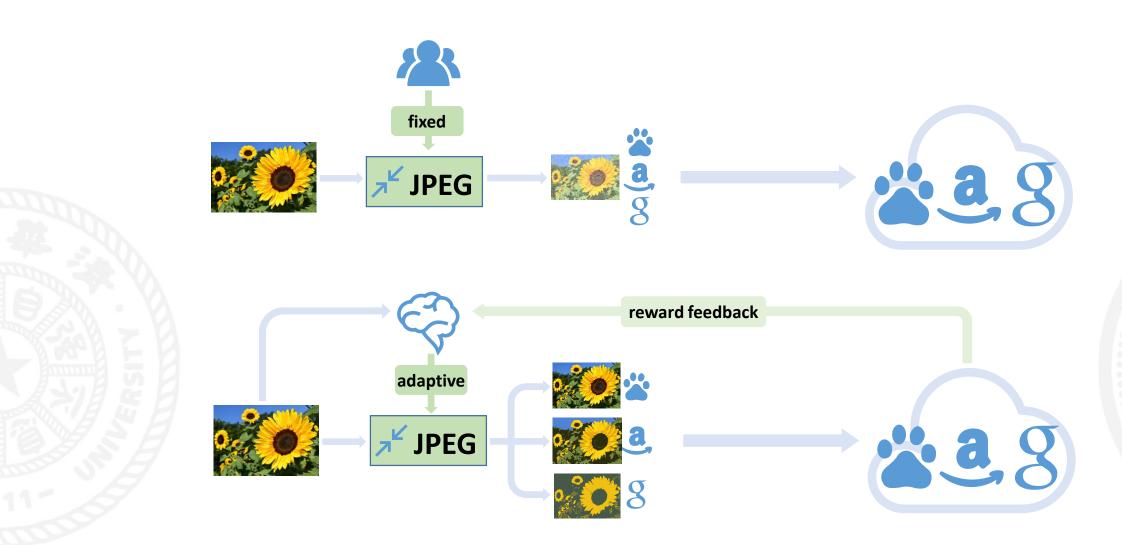




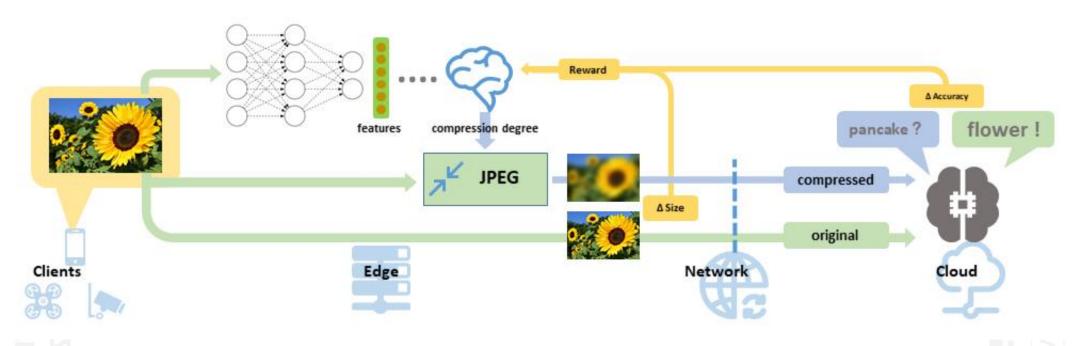
Pre-knowledge of original model





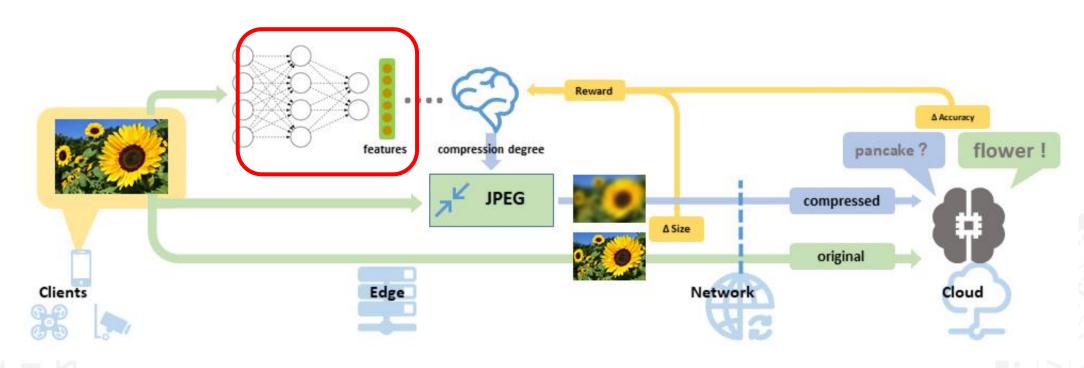






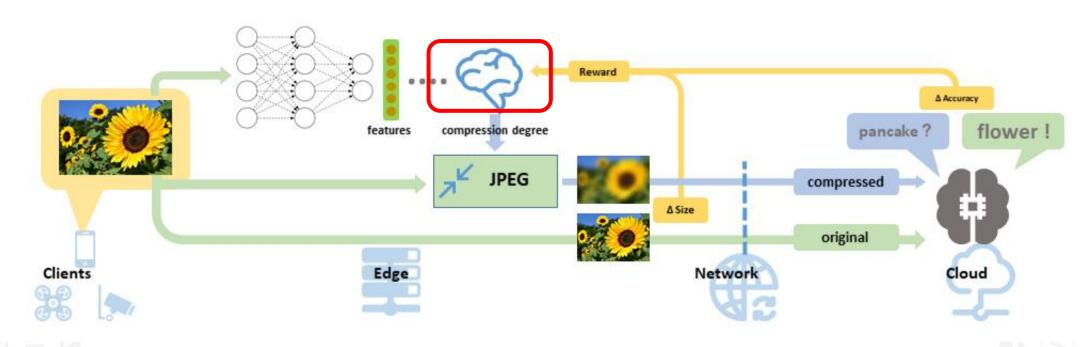
- Reward
 - Δsize Δaccuracy
- States
 - Features of the input images
- Actions
 - 10 discrete compression levels





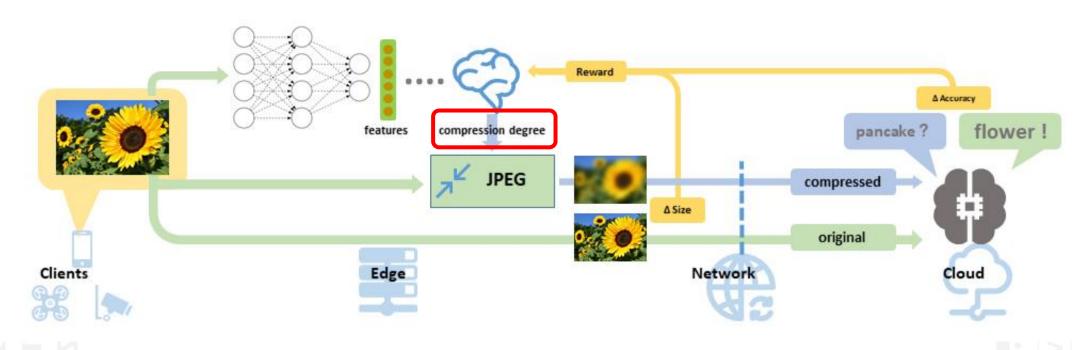
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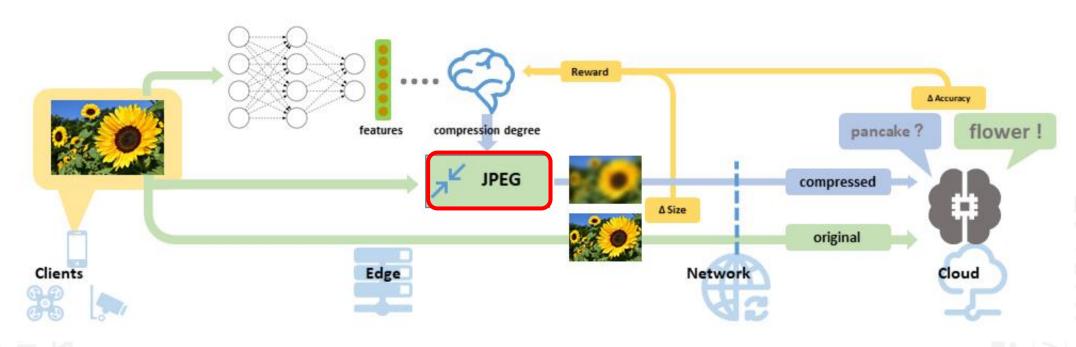
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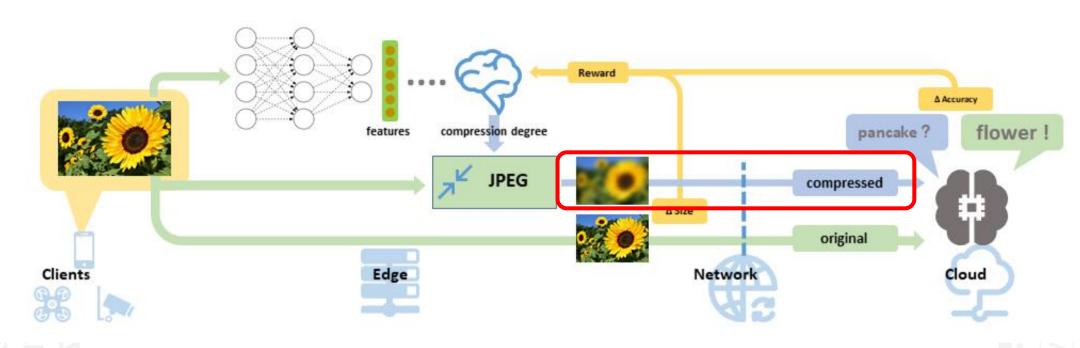
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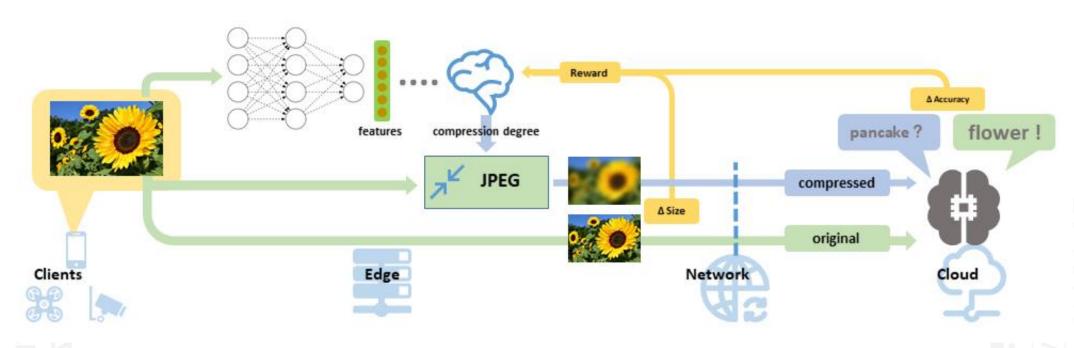
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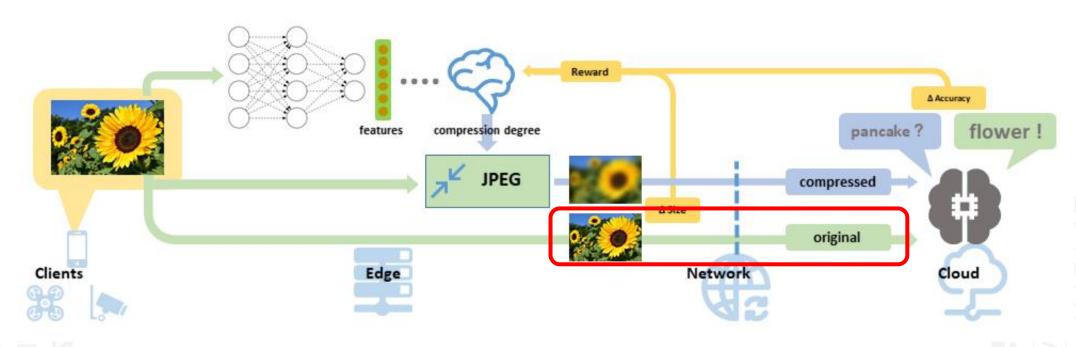
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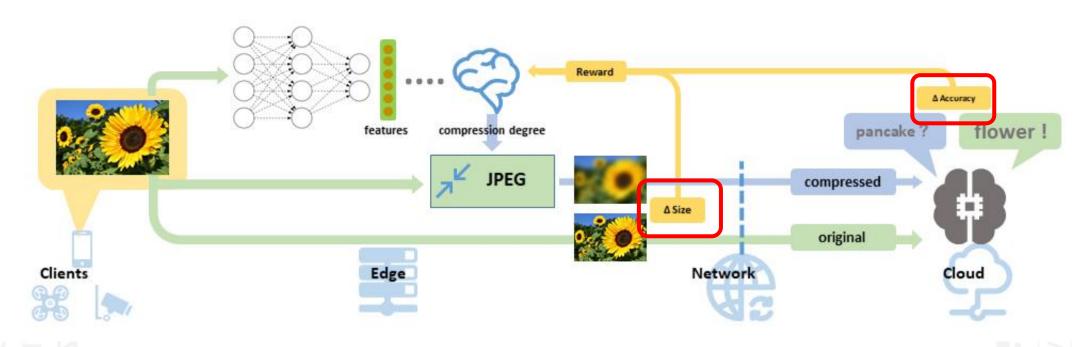
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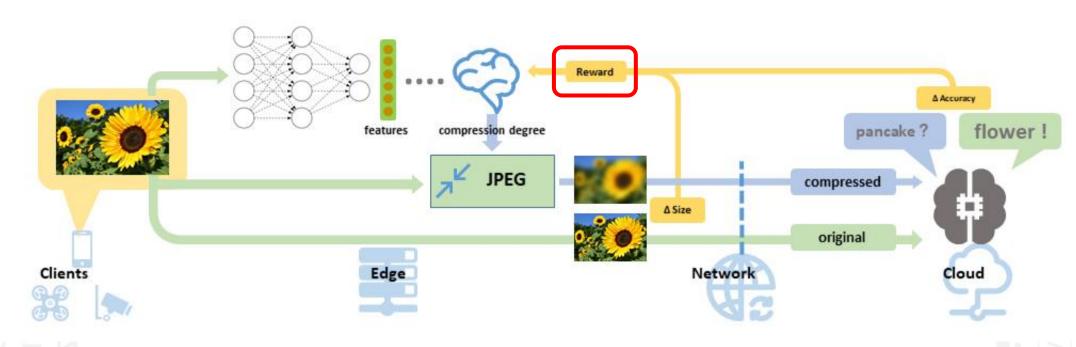
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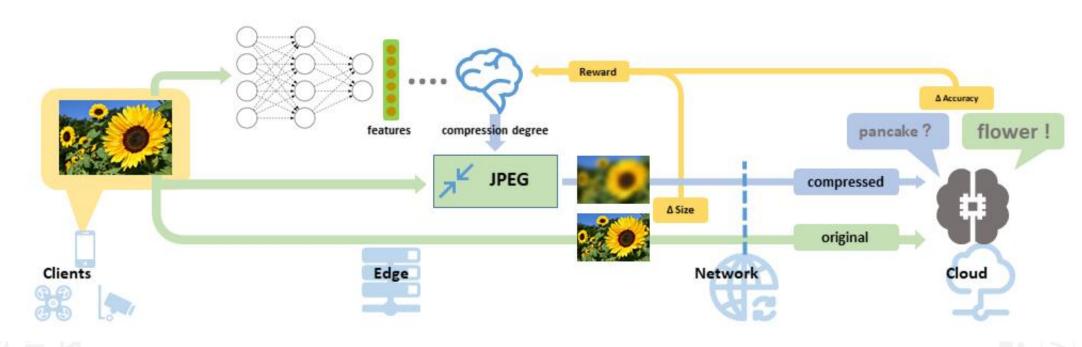
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Performance



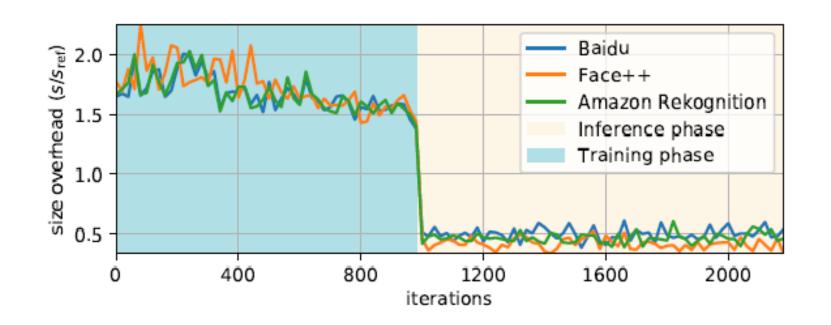


Figure 8: Upload size overhead in training and inference phase

- Size overhead in training phase
- Inference phase is longer than training phase

Performance



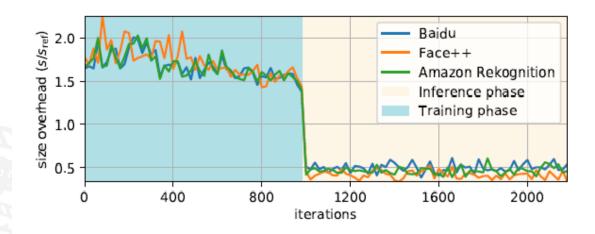


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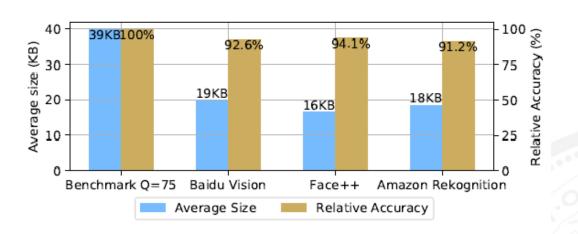


Figure 9: Average size and relative accuracy on different cloud services

- Size overhead in training phase
- Inference phase is longer than training phase

Insight



- Different compression strategies in different environments

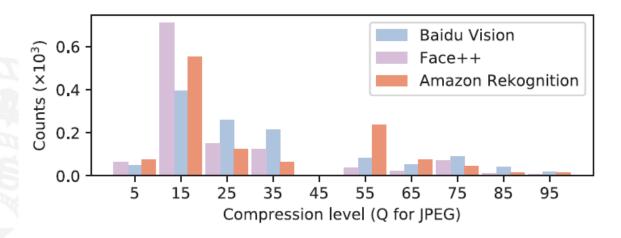


Figure 5: Histogram of RL agent's best compression level selection for different cloud services

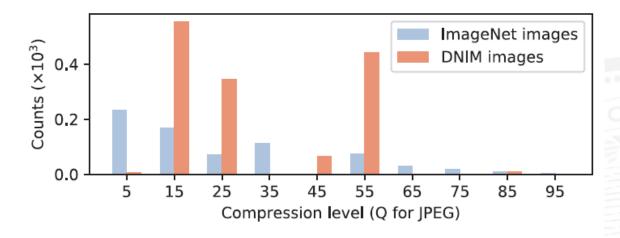


Figure 6: Histogram of RL agent's best compression level selection for different scenery image inputs

Insight



- Grad-Cam
- High quality for smooth region

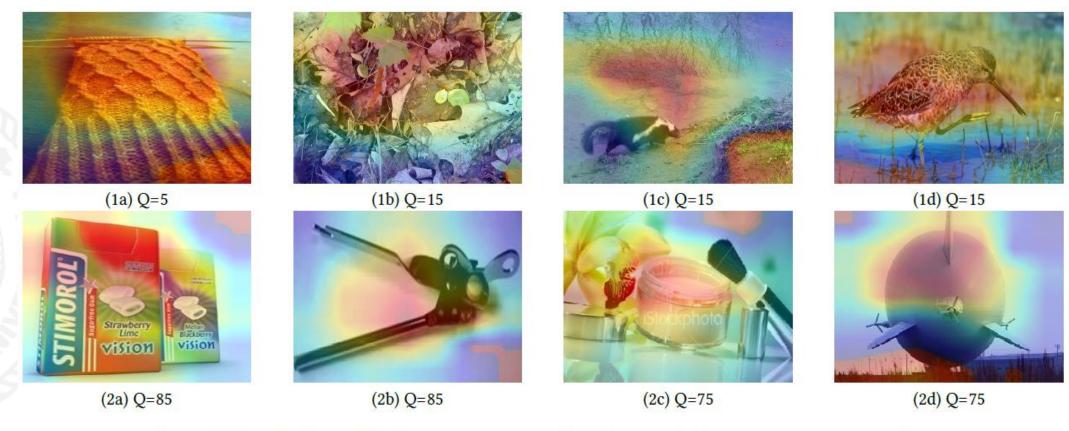


Figure 7: Visualization of the importance map for the RL agent to choose a compression quality

Scenery change



- Scenery change (day to night, sunny to rainy etc.)
- State machine with 3 states
- Occasionally estimate system accuracy
- Retrain when necessary

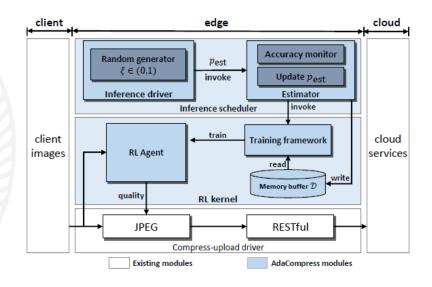


Figure 3: Diagram of AdaCompress architecture

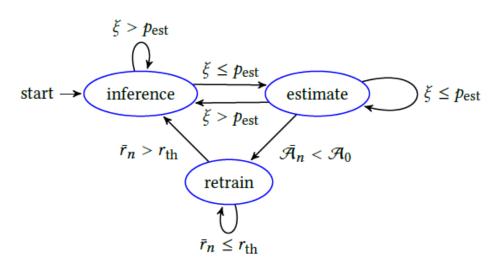


Figure 4: State switching policy



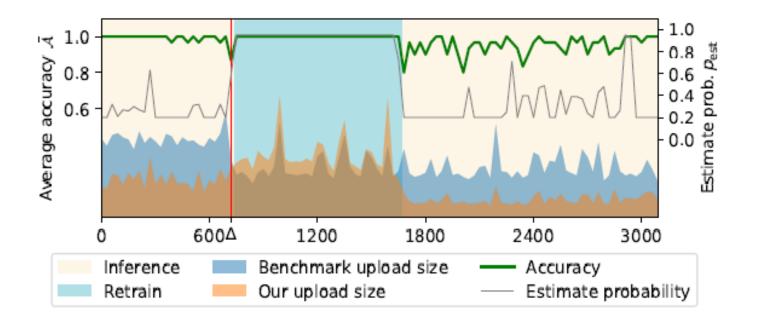


Figure 10: AdaCompress's reaction upon scenery change



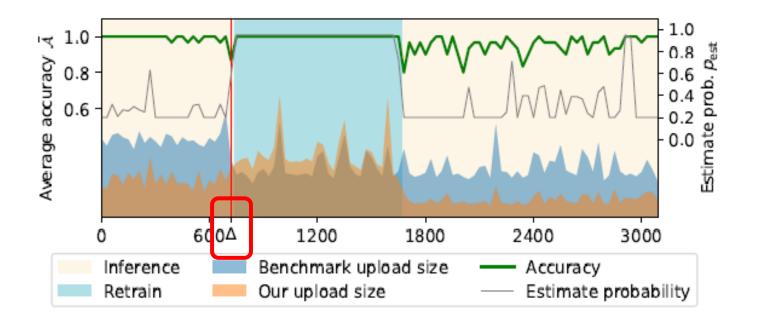


Figure 10: AdaCompress's reaction upon scenery change



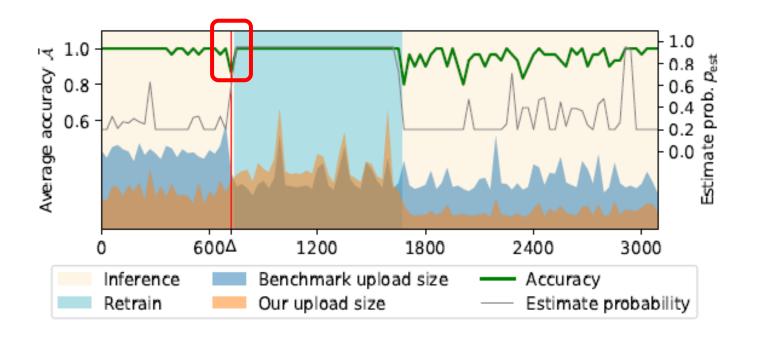


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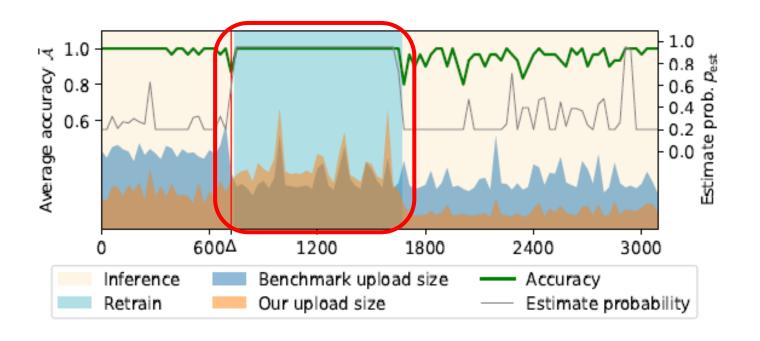


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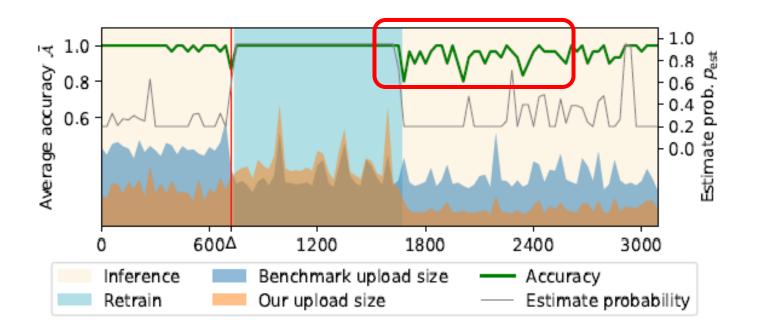


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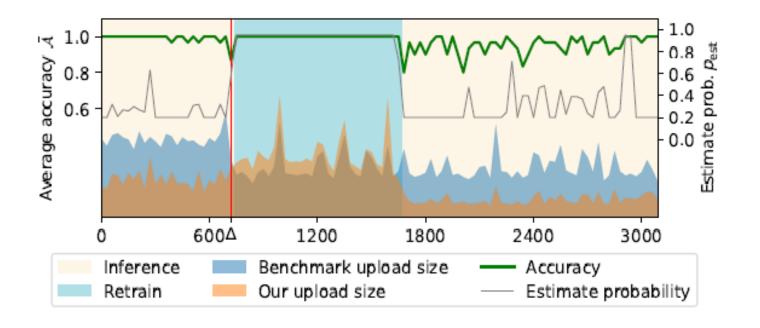


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Thank You

Source Code:

https://github.com/hosea1008/AdaCompress

