Doing Good Research and Experiments!

Dual-Degree orientation for CVIT Group https://tinyurl.com/2023CVIT-DDWorkshop

May 27th, 2023

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Doing (Good) Research

- Welcome to the CV community! <u>Good Citizen workshop @ CVPR 2018</u>
- What is Research?
- Ways to contribute
- Picking an area (work with your advisors on this)
- Publications
- On high standards
- Work ethic
- Ethics

Best Practices in ML

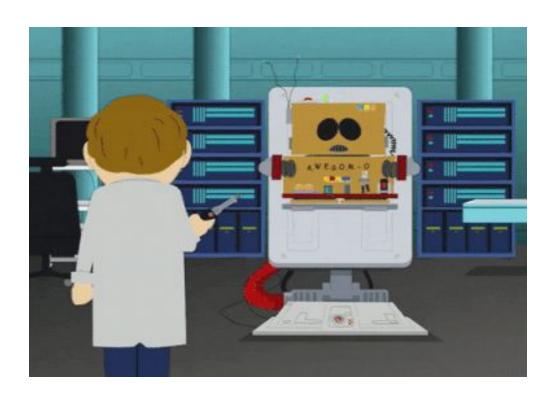
Link to the document

- Research Attitude
 - This is not coursework
 - Creating new information is not easy
 - Celebrate the small wins
 - Adopt the scientific method
- Never forget ML 101 (even when you are experienced)
 - When working with a new dataset
 - When working with a new model
 - Loss functions and plots
 - Closing the loop
- [check out the doc]



Experiments









Doing research with your advisor / mentor / collaborators

- How to work with your advisor(s)?
- How to meet with your advisors/mentors?
- How to share your progress with your mentors/collaborators?
- How to do research with my mentors effectively?

Writing well

- How to write clear and concise sentences?
- How to write a paper that looks like a good one?
- How to create a good table?



- The first pass: Quick scan
- The second pass: Figures and Tables
- The third pass: Re-implement or understand link between code and math

In general

- Read critically, understand strengths and weaknesses (unstated assumptions)
- Think of what could be improved
- Learn not just the content, but also

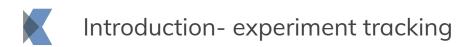
Tracking experiments with **W**and

B.Tech@IIIT-H orientation for CVIT Lab

May 27th, 2023

Agenda

Introduce the audience to WandB and its key features, to explain the benefits of experiment tracking in machine learning, and to show you how to use WandB to improve your workflow.



- What is WandB?
- Let's see some **code**
- ----- BREAK -----
- Overview of WandB sweep
- Overview, applications and functionality in WandB website

Introduction

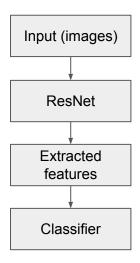
Experiment tracking and WandB.

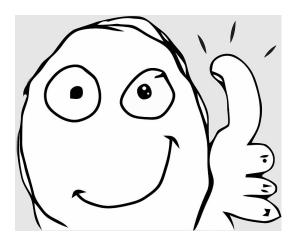


Improve your machine learning workflow

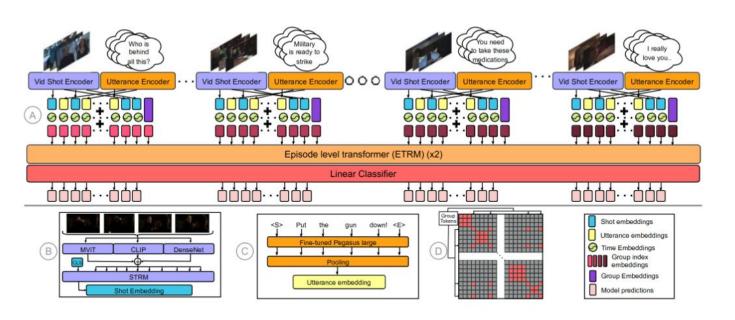
- Experiment tracking:
 - Experiment tracking refers to the process of systematically logging and organizing experiments in order to better understand the outcomes of different machine learning models and algorithms.
- Importance of experiment tracking in machine learning
 Experiment tracking is critical for machine learning practitioners as it enables them to keep track
 of their work and reproduce their results. This is particularly important in complex projects where
 multiple people are working on different parts of the codebase.

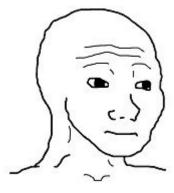
Projects grow and get complex over time.





Projects grow and get complex over time.





You may put a lot of efforts, those efforts should not be useless!







USELESS



How experiment runs usually look like

There is no fun in analysing this!

- {'epoch': 1, 'train_loss': 1.1452948740124702, 'eval_loss': 0.7713702845573426, 'train_ap_score': 0.6731415009678289,
 - 'eval_ap_score': 0.8116030823185824, 'lr': 0.0001}
- {'epoch': 2, 'train_loss': 0.5890168227255345, 'eval_loss': 0.6724603283405304, 'train_ap_score': 0.8755422621356335,
 - 'eval_ap_score': 0.8489012873912365, 'lr': 0.0001}
- {'epoch': 3, 'train_loss': 0.3640854485332966, 'eval_loss': 0.6902201867103577, 'train_ap_score': 0.9443318746126932,
 - 'eval_ap_score': 0.8558312675174203, 'lr': 0.0001}
- {'epoch': 4, 'train_loss': 0.20835940316319465, 'eval_loss': 0.733428498506546, 'train_ap_score': 0.9796843642719427,
 - 'eval_ap_score': 0.8559268354265821, 'lr': 0.0001}
- {'epoch': 5, 'train_loss': 0.1133101735264063, 'eval_loss': 0.8106619369983673, 'train_ap_score': 0.9937608794513503,
 - 'eval_ap_score': 0.8549025780413813, 'lr': 0.0001}
- $\bullet \quad \{ \text{'epoch': 6, 'train_loss': 0.06395852622576058, 'eval_loss': 0.872140085697174, 'train_ap_score': 0.9981744515435429, 'eval_loss': 0.872140085697174, 'train_ap_score': 0.872140085697174, 'train_ap_score':$
 - 'eval_ap_score': 0.856293925063724, 'lr': 0.0001}

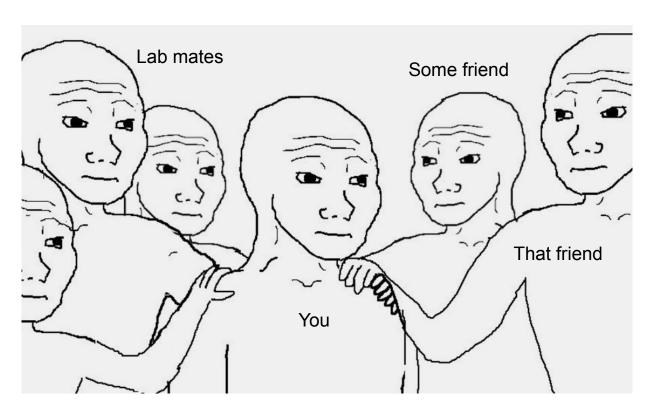


Improve your machine learning workflow

Without visualization/tracking, it is tough to answer queries like-

- How long does it take to run your experiments?
- Around what epoch does it start to overfit?
- When scheduler updated the learning rate, how much did that affect the metrics? Did it even trigger?!
- How to compare X different runs that only have change in one parameter? Which one to choose?
- Many more.....

Research work needs support



We will share one more friend with you today!



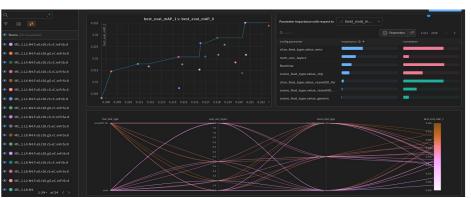


One place for all you experiments

- WandB is a powerful experiment tracking tool that helps machine learning practitioners to keep track of their models, datasets, and experiments. WandB offers a range of features including real-time visualization, hyperparameter tuning, and experiment comparison.
- Code \rightarrow github | Exp. \rightarrow WandB
- Easy to configure and use!







Lets jump to code

Log some configs and experiment data to **WandB**.

Head over to

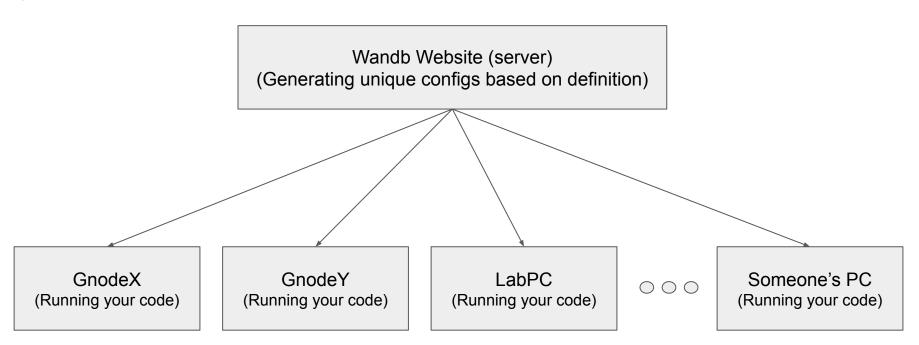
https://tinyurl.com/KathaAiWandB

https://wandb.ai/dhruv_sri/wandb_demo/

WandB sweeps



Overview



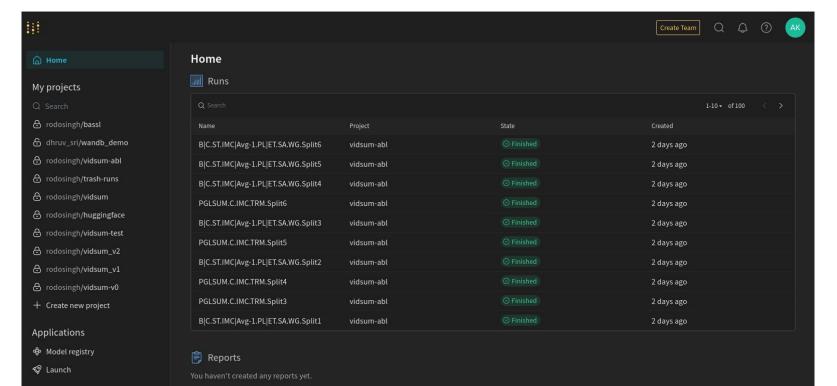
Break

Visualization is worth a thousand words : UI/UX of WandB

WandB Functionality

UI elements from COARSE to FINE and how to effectively use them

WandB Home - the GLOBAL view of all your experiments...

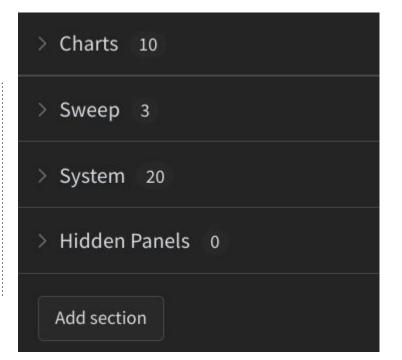


WandB Functionality

UI elements from COARSE to FINE

Sections

- Charts
- Sweep (If sweeping is done)
- System
- Hidden Panels
- Miscellaneous (Smoothing, Outlier-handling, Layout Thingy)





WandB Functionality

UI elements from **COARSE** to **FINE**

Adding a custom panel:

- Line/Bar/Scatter/Parallel Plots
- Run Comparer
- Parameter Importance
- Code
- Markdown
- Custom chart
- Weave (to check/verify project essentials)



∠ Line plot

■ Bar chart

Scatter plot

Parallel coordinates

10 Scalar chart

 $rac{N}{\mathbb{B}}$ Run comparer

FF Parameter importance

Code

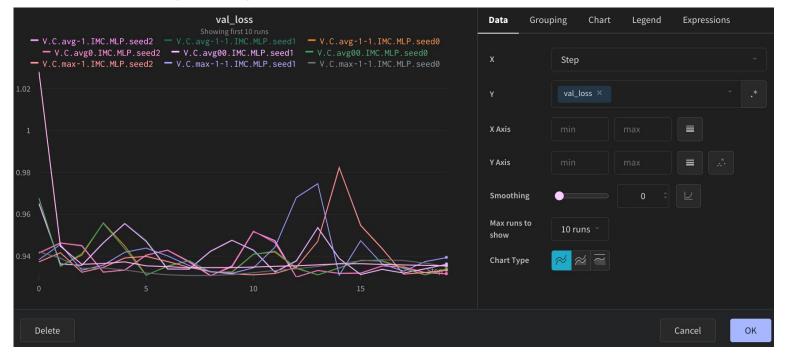
 ${\mathbb T}$ Markdown

Custom chart

Q Weave



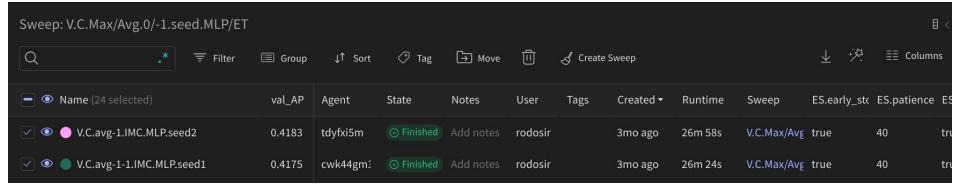
- Edit panel (Data, Grouping,
- Chart, Legend, Expression)
- Miscellaneous (Export, Duplicate, Download)



WandB RUNs - Table View



- Search / Filter / Group
- Sort
- Taq
- Move / Delete
- Magic wand (feature selection)
- Columns







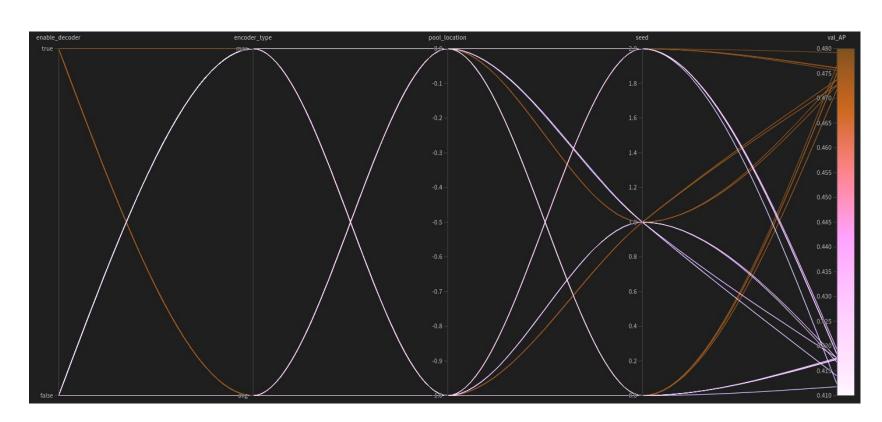
Generating configs automatically

- <u>Efficient ablation management</u>
- Grid Search
- <u>Visualizing and Comparing</u> multiple runs (of our choice) through <u>parallel</u> <u>coordinates</u> plot.
- We can engage as many gnodes (agents/workers) we want just by mentioning sweep-id.

WandB Parallel Plots - A step towards HyperOpt



Parallel plots - OUR SAVIOR



Application of Sweeps (Revisited)



- <u>Efficient ablation management</u>: Hassle free auto execution of every combination of varying parameters.
- Grid Search: Can estimate feature importance (can be selected manually) based on metric objective.
- <u>Visualizing and Comparing</u> multiple runs (of our choice) through <u>parallel coordinates</u> plot.
- We can engage as many gnodes (agents/workers) we want just by mentioning sweep-id while submitting batch job for each gnode. This parallely completes all the runs expected in that sweep. No need to submit usual batch job (with all pre-requirements of data and code satisfied for that anode).
- So it does have a lot of perks, Agree?

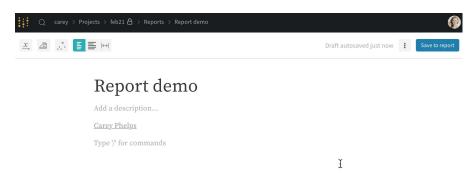








WandB Reports - [Better way to share your findings]

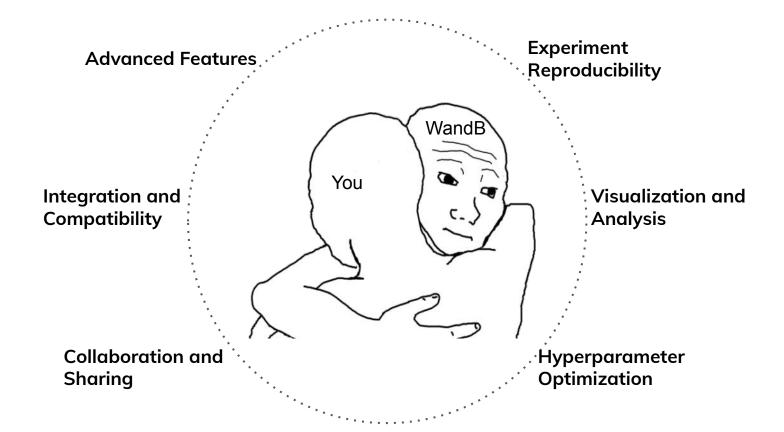


References:

- 1. <u>Link1</u>
- 2. <u>Link2</u>
- 3. <u>Link3</u>

ACHIEVE - 6 Perks of WandB









Good Research practices

- Find the problem not the solution.
- Thorough literature review
- Designing experiments
- Analysing the outcomes
- Writing a paper
- Submitting to a conference
- Releasing the code and dataset
- Publicize your work on social media!

Multiprocessing









Single Processing:(

Concurrency

Multiprocessing





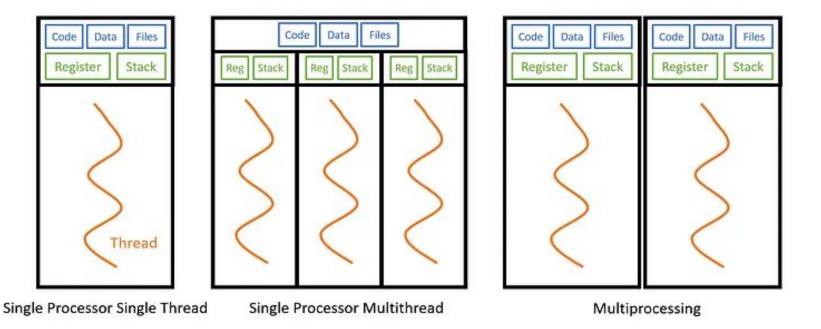


Image Credits: Silberschatz et al., "Operating System Concepts", Chapter 4

Multimodality





Credits: https://www.dreamstime.com/royalty-free-stock-photos-juggling-color-balls-acrobat-performer-image17323338





 VLMs are a type of neural network models that combines computer vision and natural language processing techniques to perform a variety of different multimodal tasks such as image captioning, visual question answering, and image retrieval.

• VLMs (like CLIP, Flamingo, BLIP, etc.) have been highly successful in recent years, with state-of-the-art results on various benchmark datasets for the above mentioned tasks.

What tasks can VLMs perform?



- Image Captioning
- Visual Question Answering (VQA)
- Visual Grounding
- Image-Text Matching
- Text to Image Retrieval
- Visual Scene Understanding

Image Captioning





A politician receives a gift from politician.



A collage of different colored ties on a white background.



Silhouette of a woman practicing Aerial view of a road in autumn. yoga on the beach at sunset.



Image Credits: https://arxiv.org/abs/2111.09734 (ClipCap)





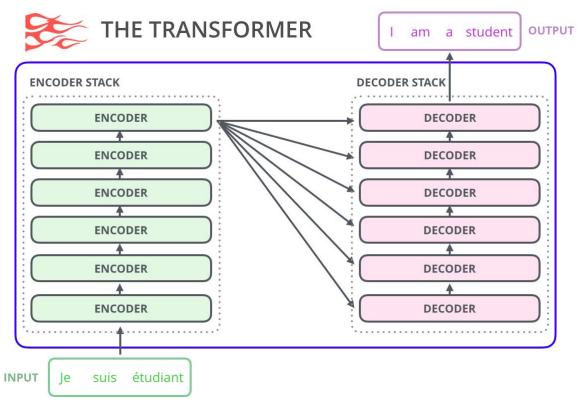


Image Credits: https://jalammar.github.io/illustrated-gpt2/



CLIP (Contrastive Language-Image Pre-training)

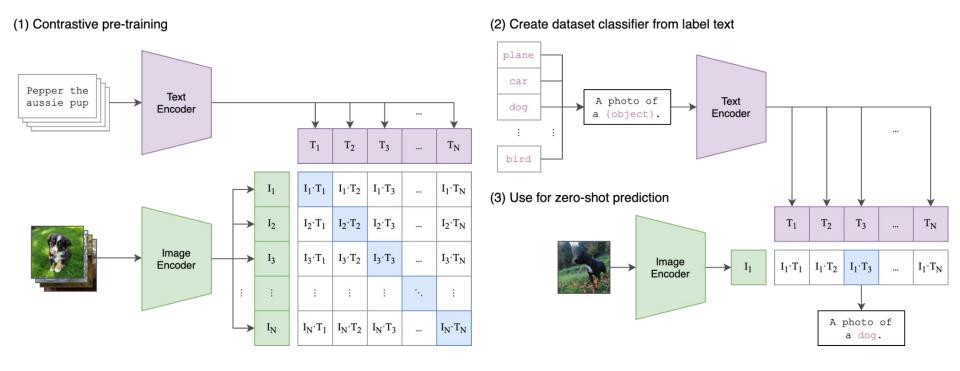


Image Credits: https://github.com/openai/CLIP







Credits: https://www.disneyclips.com/images/images/genie-lamp2.png

GPT-2



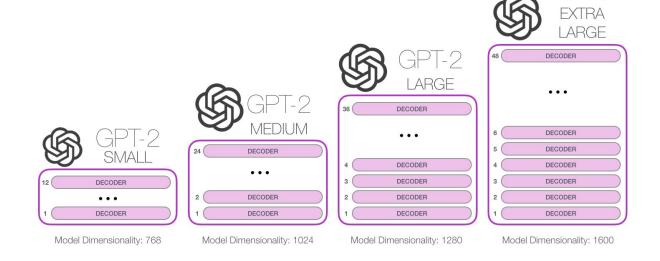
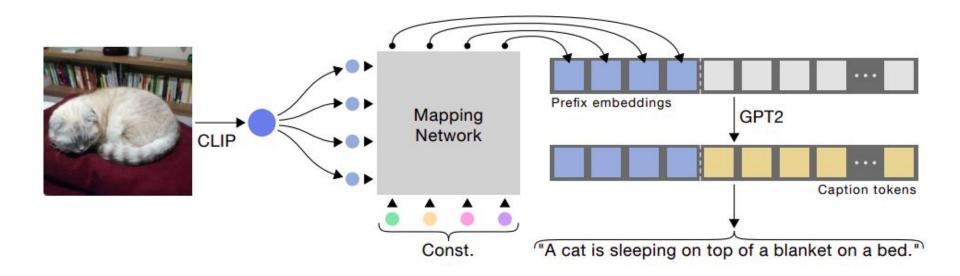


Image Credits: https://jalammar.github.io/illustrated-gpt2/







ClipCap: Objective Function¹



Dataset: images and captions $\{x^i, c^i\}_{i=1}^N$

$$c^i = c^i_1, \dots, c^i_\ell$$

$$p_1^i, \ldots, p_k^i = F(CLIP(x^i))$$

$$\mathcal{L}_X = -\sum_{i=1}^{N} \sum_{j=1}^{\ell} \log p_{\theta}(c_j^i | p_1^i, \dots, p_k^i, c_1^i, \dots, c_{j-1}^i)$$



How and where to train the models?





Credits: https://xprilion.com/ml-on-2gb-ram/

ClipCap¹



Colab Notebook: tinyurl.com/mu8u4z9y

(https://colab.research.google.com/drive/1ligZ0ZUABgcFWHKPmTgK6Sa_SVXB7hu3?usp=sharing)

