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Embedding Domain Knowledge
Al shot

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e case

The Learning Spectrum

Semi-Supervised Learning

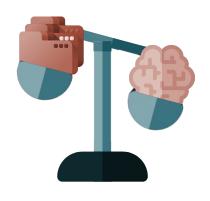
01

The good old days



Before and After Deep Learning





- Tons of focus on feature engineering and domain understanding.
- Manual-intensive task.
- Do you need a COVID-19 detection model?
 - Sure! give me 10 Ph.D., a budget for 5 years, 1K images.
 - Then, I'll come close to a PoC that can only be used under these strictly designed settings.



Before and After Deep Learning

After Deep Learning



- The new popular kid arrived to the school.
- Getting models for new applications "for-free".
- Do you need a COVID-19 detection model?
 - Sure! give me 1 month, a monkey capable of pressing two buttons (fit+predict) and 1M data points.
 - Then, I'll have a model that seems to be working better than humans... at least on silico.



Before and After Deep Learning



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- So, people got mad...



Before and After Deep Learning



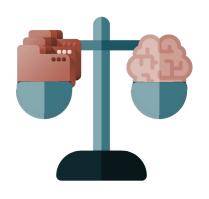
After Deep Learning

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 - Then, I'll have a model that seems to be working better than humans... at least on silico.
- So, people got mad... up to the point where we almost discard human experience as a valid source of knowledge.



Before and After Deep Learning

Embedding Domain Knowledge in DNN



- Not so bald nor with two wigs
 - o (a bite of venezuelan wisdom)

- We can have a mid-point where:
 - We understand the business.
 - We understand the tech.
 - We know how to elevate the tech to the business.
 - Instead of dumbing down the business to fit the tech.



02 Embedding Domain Knowledge in DNN



We will discuss multiple examples and how to build tailored DNN for each case...

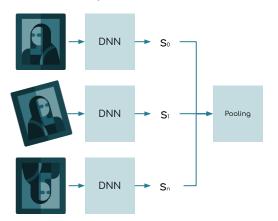
Fasten your seatbelts. Please try this at home!



Adding Invariances: The easy way

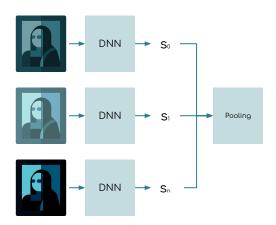
Synthetic (or natural) Data Augmentation at Inference



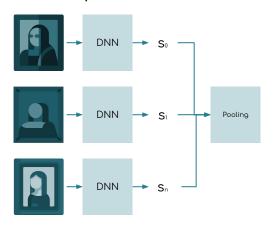




Illumination/Contrast



Multiple Views/Poses





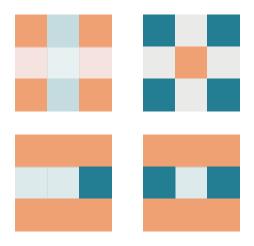
Adding Invariances: The hard way

Local Preprocessing or...

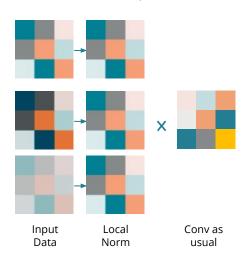
Building tailored convolutional kernels/architectures



Flips/Rotations



Illumination/Contrast



Multiple Views/Poses

Ask Geoffrey Hinton about capsule neural networks



Adding Invariances

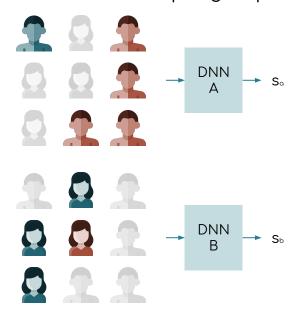
Group invariance:

race, gender, country, deep-learning-framework-preference

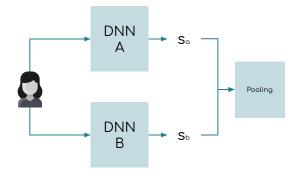
Biased Dataset



Train a model per group



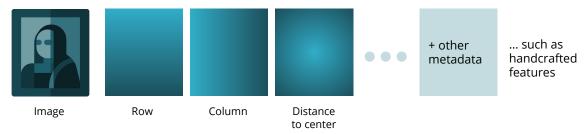
Avg. voting as prediction

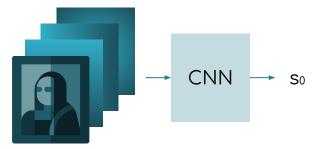




Removing Invariances

- Remove translational invariance from convolutional kernels
 - Add pixel location as additional channels



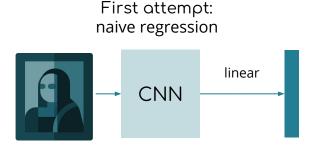


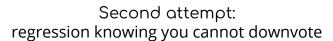
Note: you can have two sets of kernels, one with and one without spatial coordinates so you get the best from the two worlds.

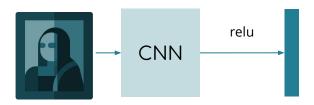


Social Media Reactions

Predict: likes, love, care, haha, wow, sad, angry



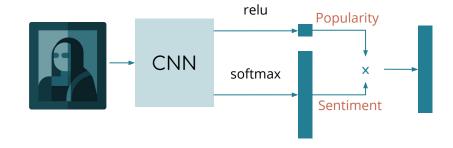




Third attempt:

Easen the task by learning 2 simpler tasks:

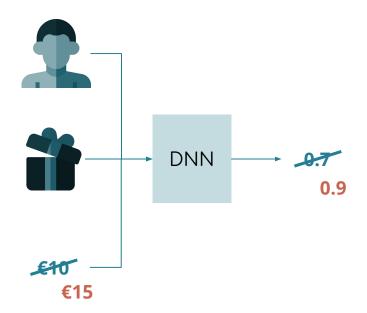
- Popularity
- Sentiment

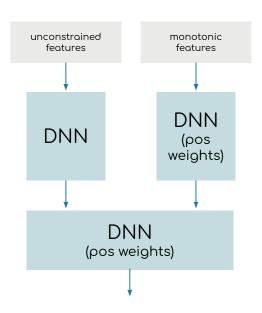




Predict if a customer will buy a product

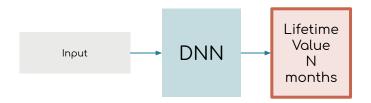
Forcing monotonic Behavior





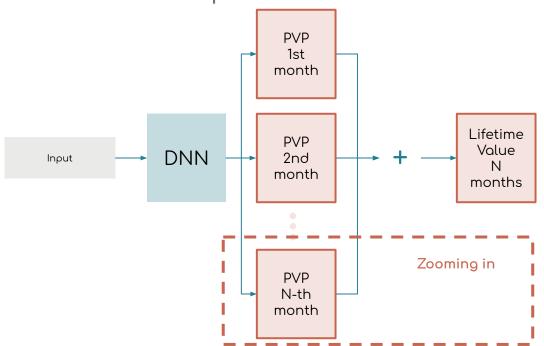


When harder is simpler





When harder is simpler



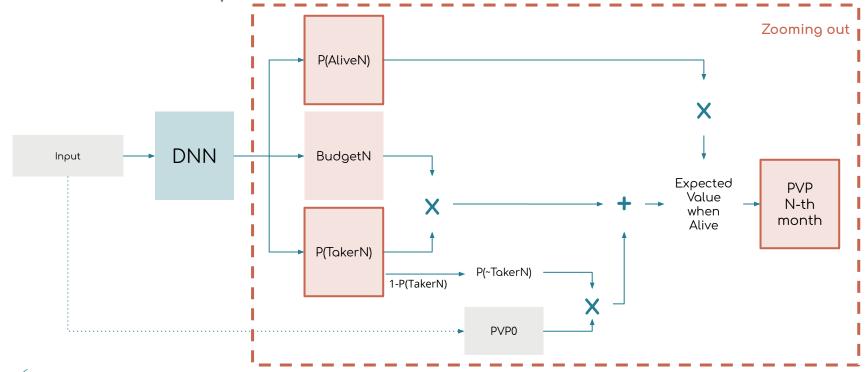
Expected PVP = Prob(Alive) *
TotalBudget + (1 - Prob(Alive)) * 0

TotalBudget = Prob(Taker) * Budget + (1 - Prob(Taker)) * InitialPVP

Expected PVP = Prob(Alive) * (Prob(Taker) * Budget + (1 - Prob(Taker)) * InitialPVP)

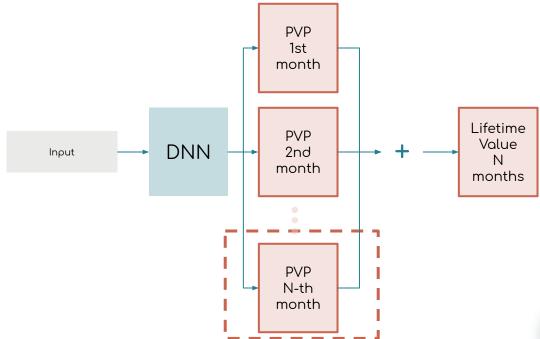


When harder is simpler





When harder is simpler



Why adding complexity and investing that much effort?

- Additional tasks act as regularizers.
- Giving additional supervision of the internal business processes:
 - Facilitates the discovery of features relevant for those tasks.
 - Avoids catastrophic failures due to unmatching business-machine rules.
- We reduced the MAE by 50%;-)



https://nilg.ai/blog/202004/embed ding-domain-knowledge-for-estim ating-customer-lifetime-value/



Conclusion



- Deep Neural Networks are not the black-box monster they told you.
- They can actually be a good dog if you know how to train (or breed) them!
- Domain knowledge isn't (always) bad.
 - It can help you to learn with **less data**.
 - It can help you avoid **catastrophic failure**.
 - So you can rest tonight! Assured that your Project
 Manager won't call you at midnight because your
 model decided to sell cars at \$0 to maximize hit rate.



Thank you! Questions?





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