

Unfit for intended use: Detecting emergent bias in machine learning systems

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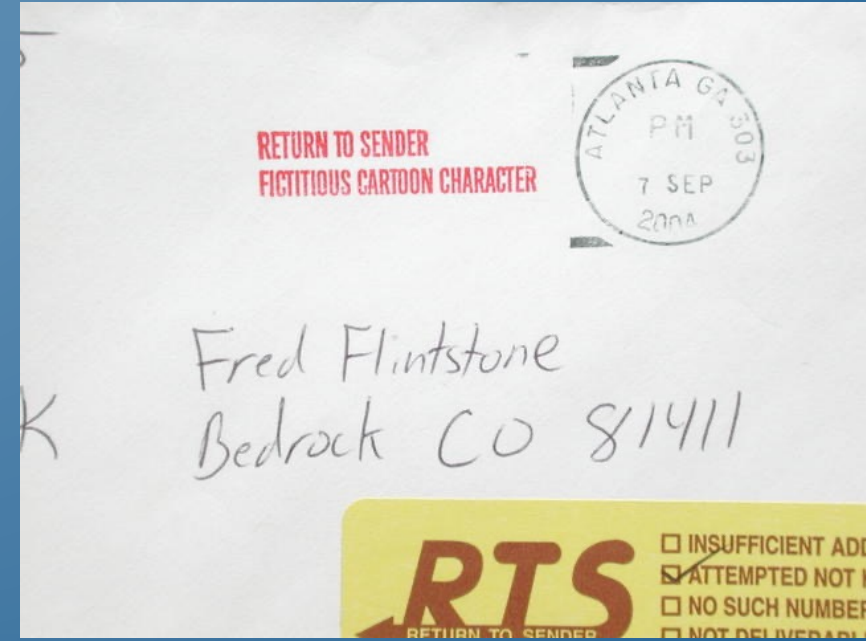
Topics for today

1. What is emergent bias for machine learning systems?
2. When does emergent bias happen?
3. How can we detect it before using/deploying the ML model?

Goal: Recognize handwritten numbers

- Examples

- U.S. ZIP Codes on postal mail
- Phone numbers
- Handwritten check amounts
- Time of day on a handwritten note
- Dosage on a medical prescription
- (Etc.)



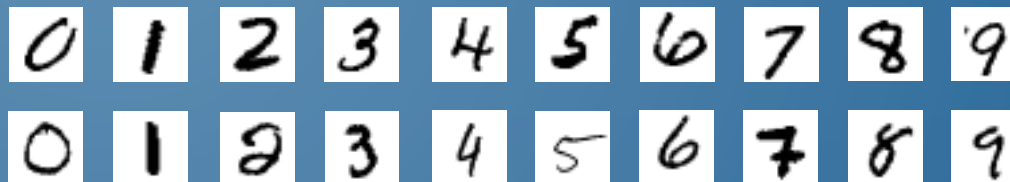
Multi-Digit Writer (MDW) U.S. ZIP Codes

- Synthetic handwritten **multi-digit** numbers written **by a single writer**

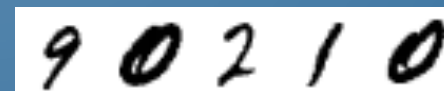
- Assembled from MNIST digits (LeCun et al., 1998)



Yann LeCun



Writer 2445



Writer 489



Writer 480



Writer 1423



Writer 3856



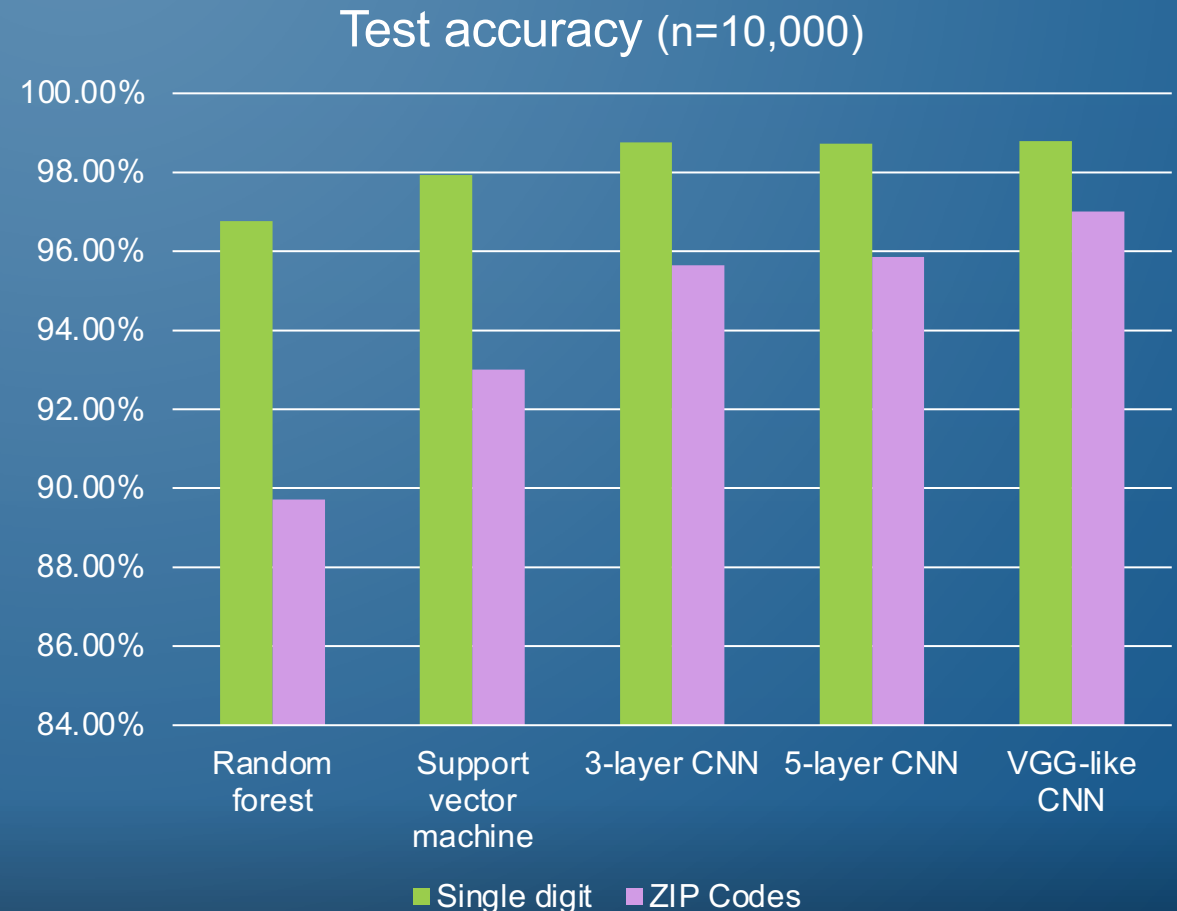
Writer 2389

- Limitations: artificially regular spacing, doesn't capture sequential writing behaviors, each digit is only 28x28 pixels
 - Prefer to collect new (sequential) handwriting samples if resources permit

Which ZIP Code model do you prefer?

- Recognizing 5-digit ZIP Codes is harder than single digits

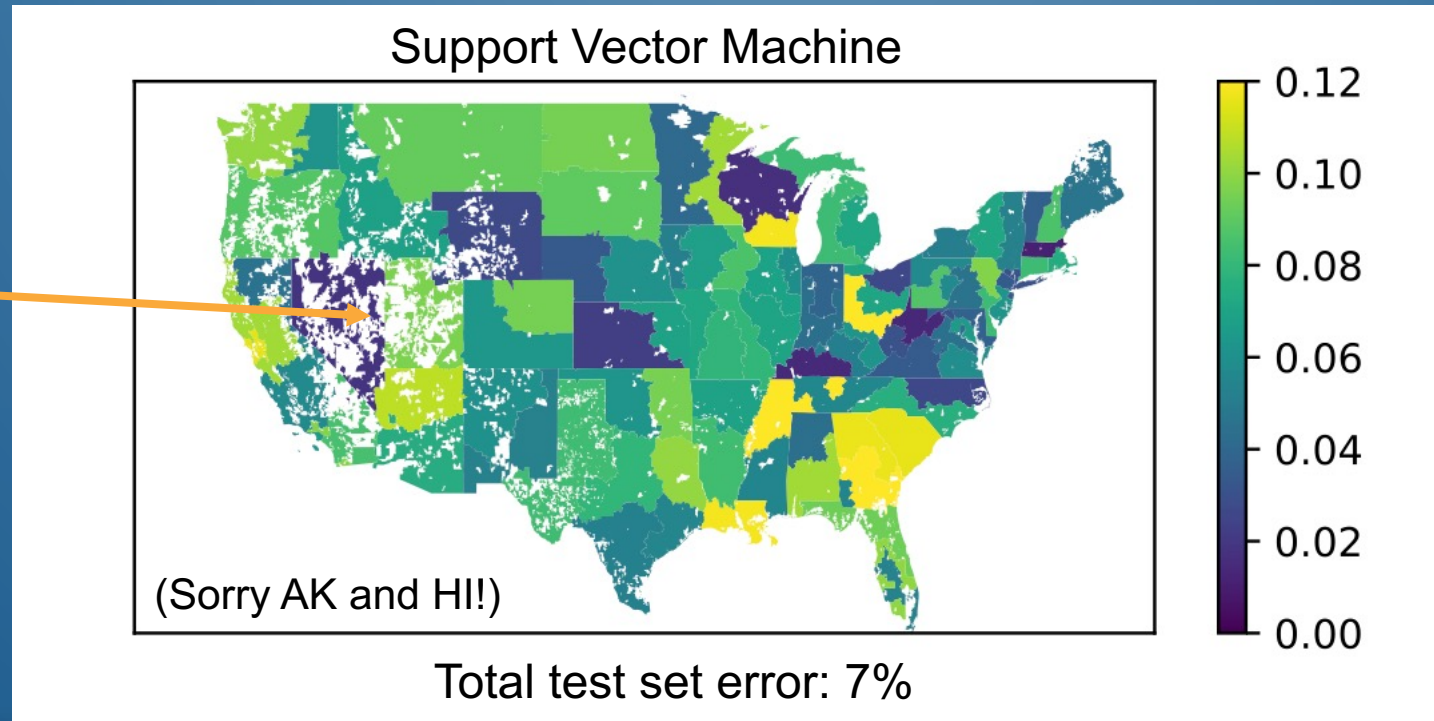
	Single digit	5 digits
Classifier (n=60,000 training items)	MNIST test accuracy (n=10,000)	MDW ZIP Codes test accuracy (n=10,000)
Random forest	96.77%	89.71%
Support vector machine	97.93%	93.00%
3-layer CNN	98.75%	95.64%
5-layer CNN	98.73%	95.85%
VGG-like CNN	98.79%	97.01%



Error is not equally distributed (geographical bias)

- Error rates (lower is better) per U.S. ZIP Code sector
 - Sector: all ZIP Codes with the same first two digits (e.g., 00XXX, 01XXX...)

Some places
do not belong
to a ZIP Code!

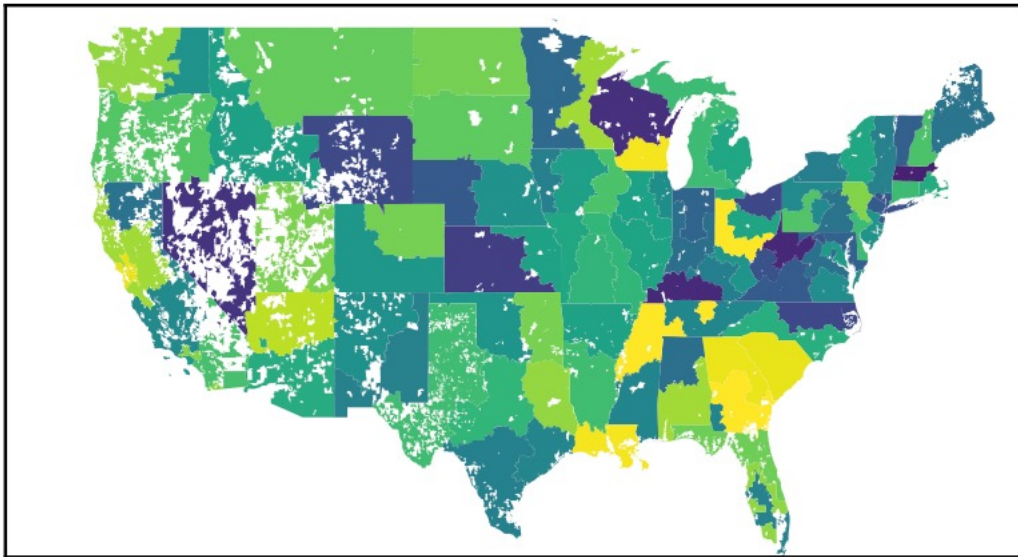


Emergent bias: performance differs by location
(Bias = differential performance for sub-groups)

Which ZIP Code model do you prefer?

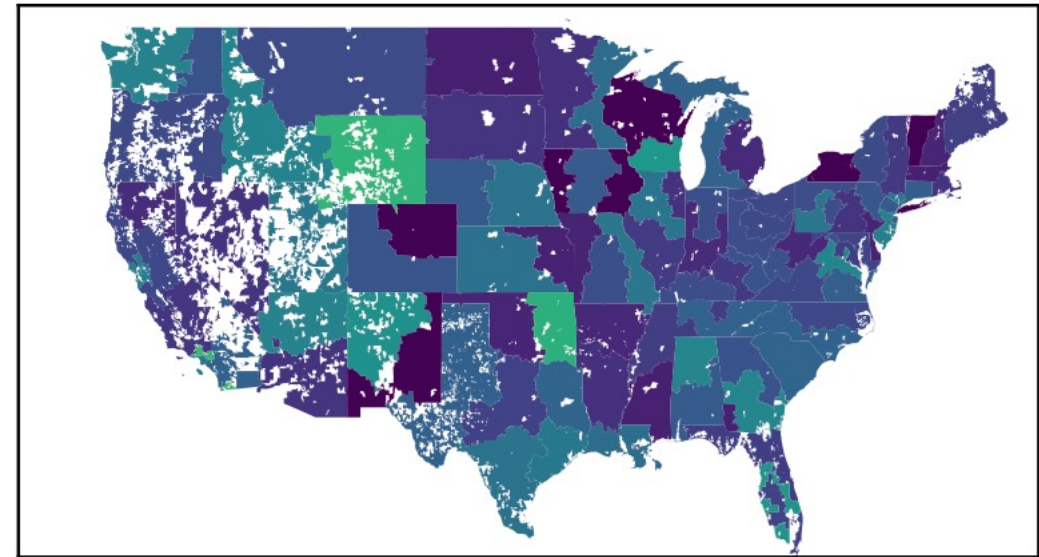
- VGG-like CNN has lower total error
 - But its emergent bias penalizes different areas than the SVM

Support Vector Machine

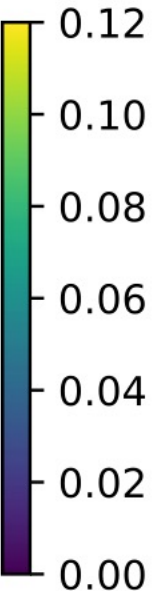


Total test set error: 7%

VGG-like Convolutional Neural Network



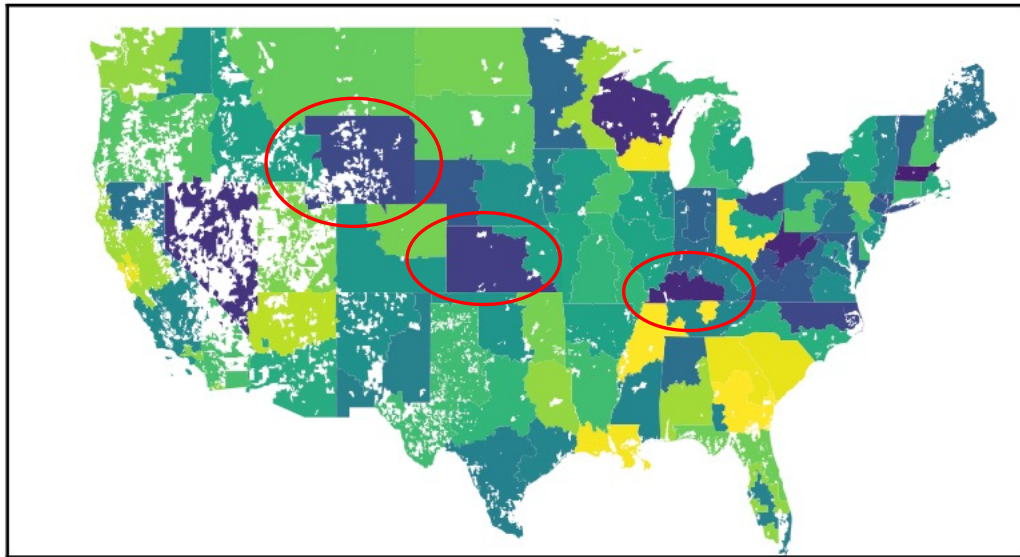
Total test set error: 3%



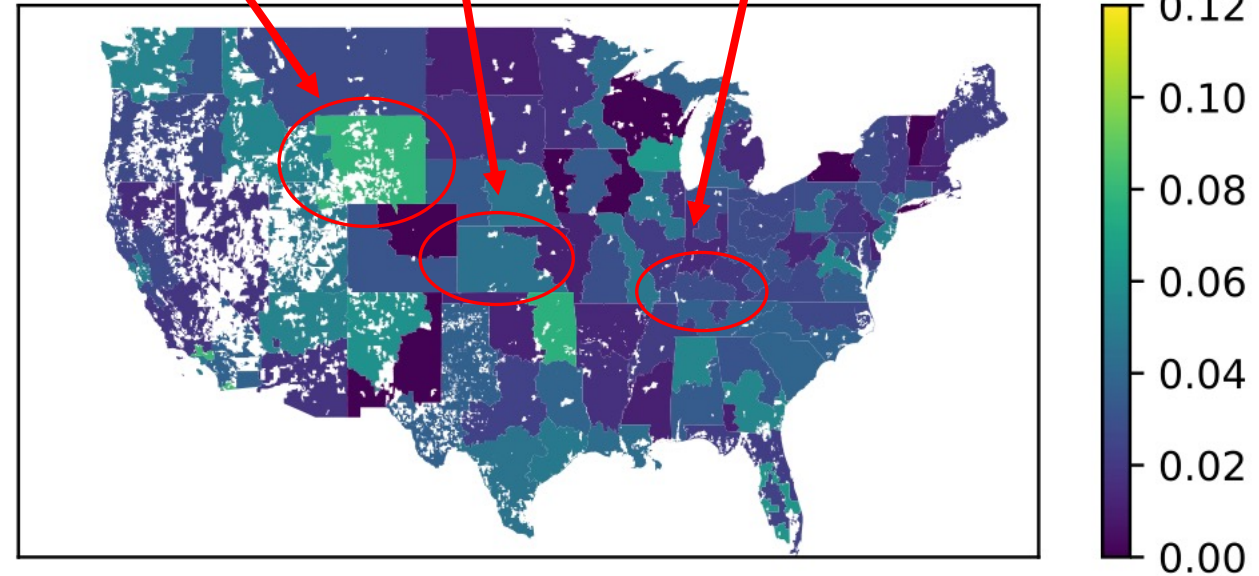
Different emergent bias in each model

What if you live in Wyoming, Kansas, or Kentucky?

Support Vector Machine

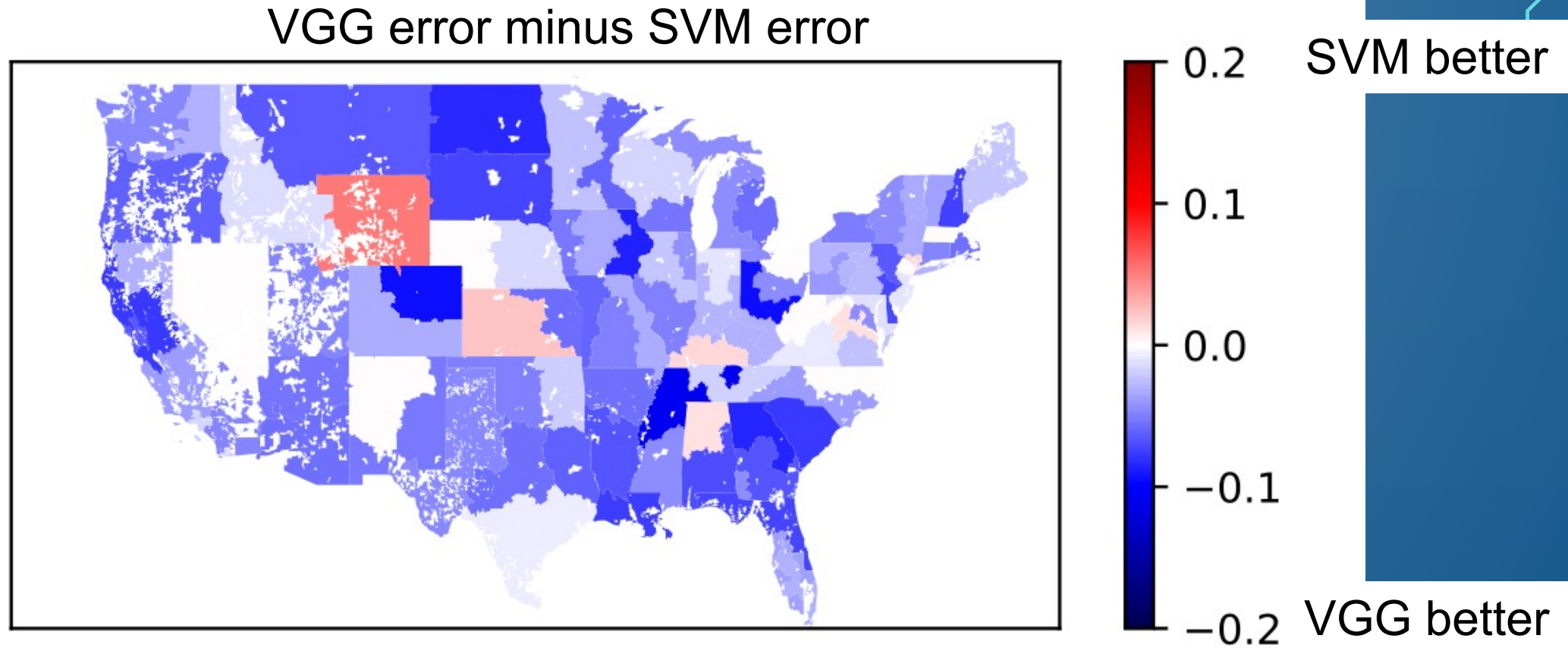


VGG-like Convolutional Neural Network



Different emergent bias in each model

Comparative geographical bias VGG and SVM models



Should we use a different model in each location?

We need to evaluate emergent bias carefully to inform deployment decisions

What is emergent bias?

- Emergent bias: Bias that manifests (only) when applying a system in contexts not originally envisioned
 - Geographical bias is not in the training data
 - Geographical bias is not even defined for isolated digit classifiers
- “While it is almost always possible to identify preexisting bias and technical bias in a system design at the time of creation or implementation, **emergent bias arises only in a context of use.**”
(Friedman & Nissenbaum, 1996)

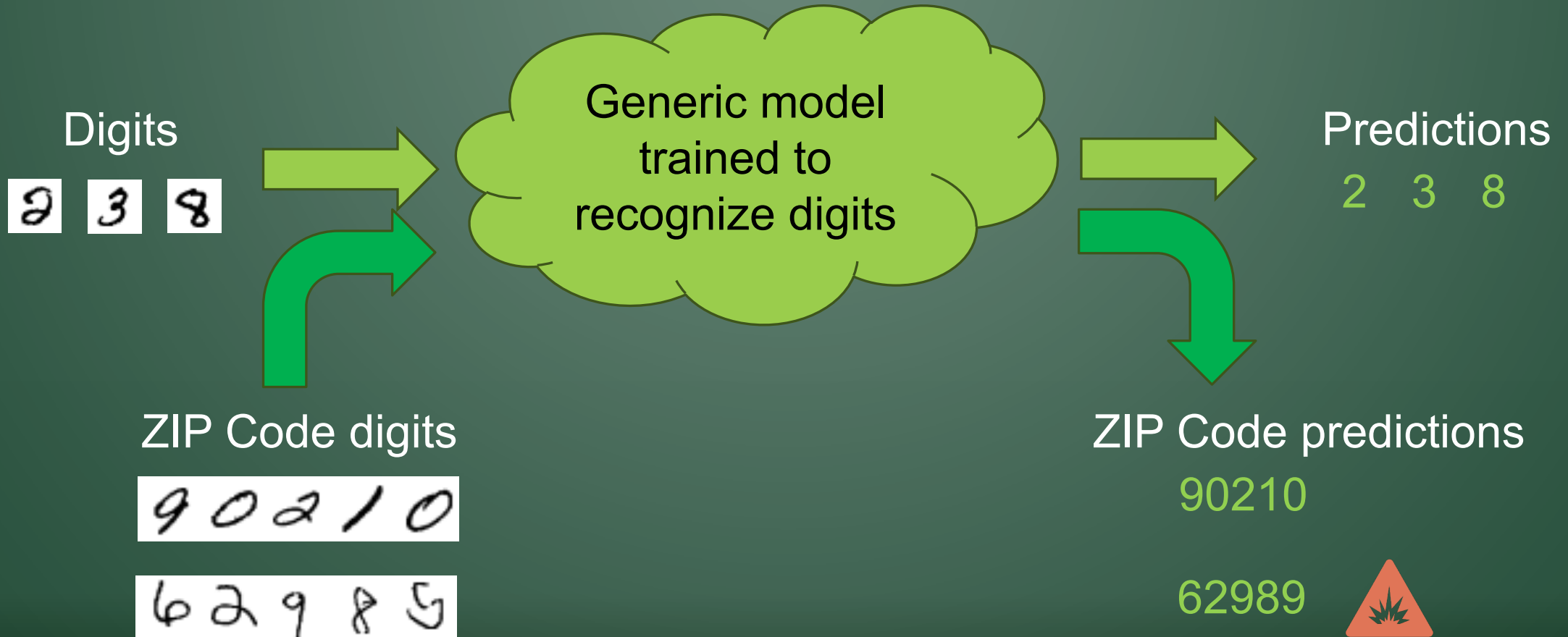


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Is the model unfit for intended use?

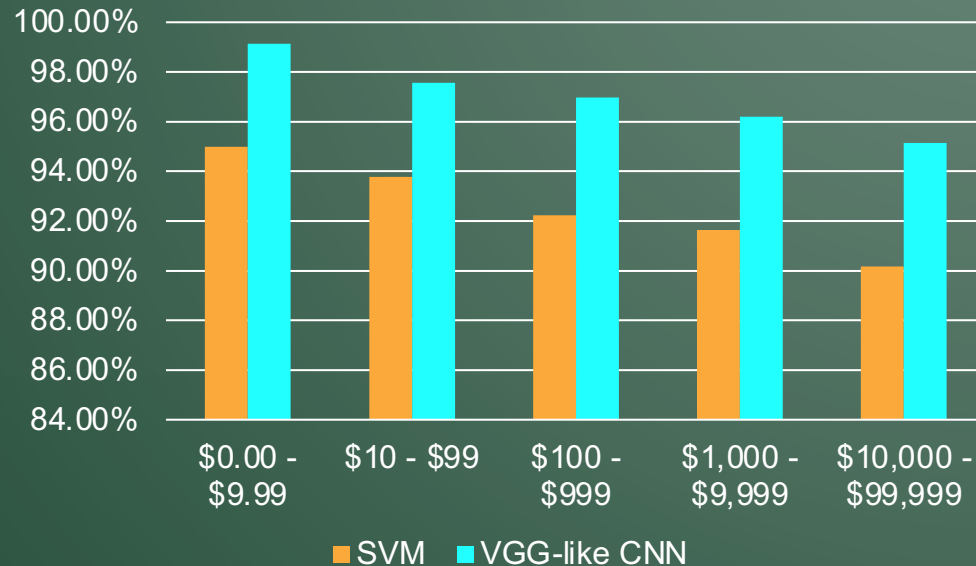
- What went wrong?



More examples

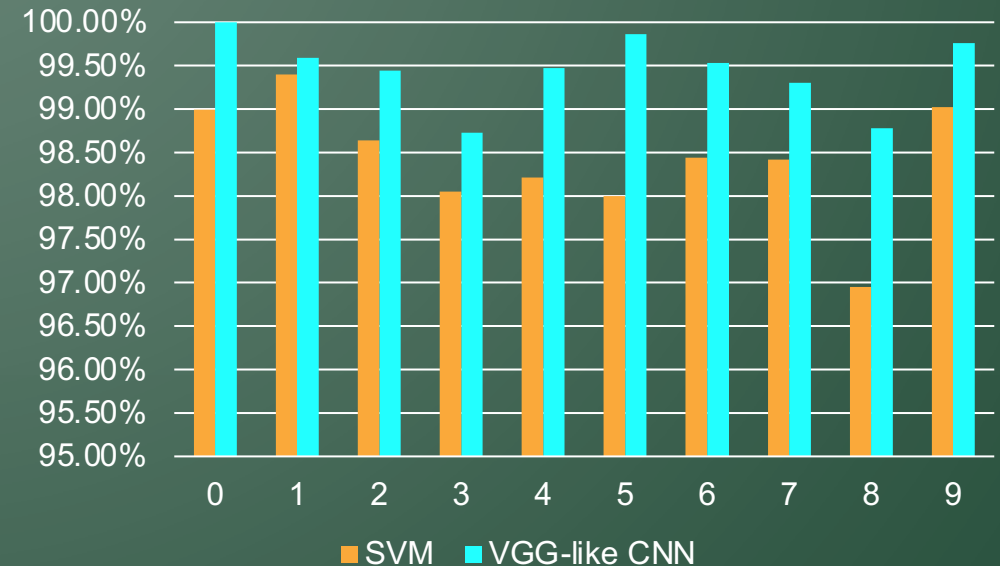
- MDW handwritten check amounts: \$0.00 to \$99,999
 - The position of each digit matters

Accuracy by **amount** (n=10,000)



**Longer numbers are harder
for both models**
(not surprising, but not what you want)

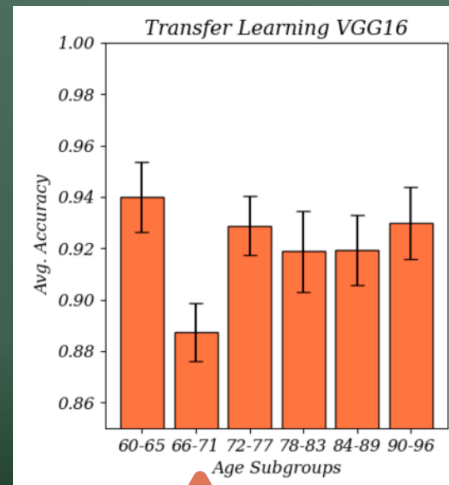
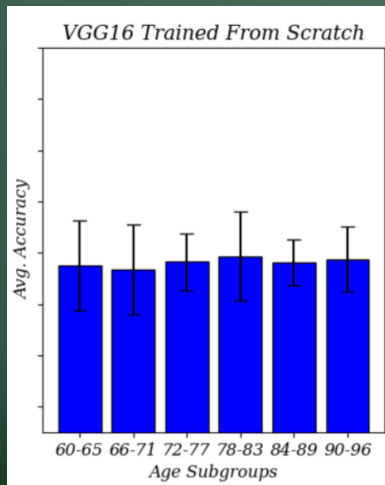
Accuracy by **leading digit** (n=10,000)



**SVM has a weakness for
amounts starting with \$8**
(surprising!)

Other causes of emergent bias

- Domain shift: Changes over time and/or space
- Transfer learning
 - Alzheimer's disease diagnosis from brain scan images (Salmani & Lewis, 2024)
 - VGG16 model trained on ImageNet and fine-tuned for brain scans exhibits 3x more bias (gender) and 5x more bias (age) compared to VGG16 model trained from scratch



Avg. performance for fine-tuned VGG16 is higher (92.3% vs. 91.7%), but it exhibits emergent bias

Detecting emergent bias before deployment

- The generic MNIST test set cannot reveal this bias to you
 - “**emergent bias arises only in a context of use.**” (Friedman & Nissenbaum, 1996)
- Do not rely on generic metrics (accuracy, AUC, F1, etc.) to do model selection
- Ask: What performance metric do you actually care about?
 - Error rate across the entire sample
 - Error rate per region
 - Error rate for urban vs. rural
 - Error rate per region, weighted by population
 - Error rate per region, weighted by postal activity

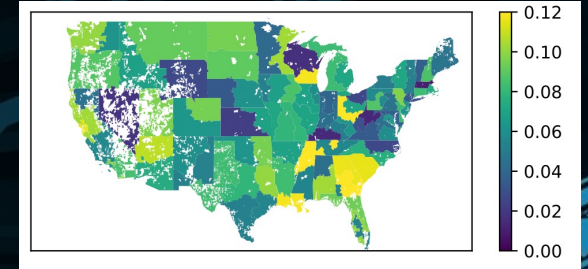
What are the sub-groups of interest?

What can you do to reduce emergent bias?

- Collect more data for sub-groups with lower performance and re-train
- Selectively deploy best classifier for each sub-group
 - But this means you are making decisions based on fewer examples per sub-group, so proceed with caution, especially for minority sub-groups

Emergent bias in machine learning systems

- Emergent bias: appears in a specific context of use
 - It is not measurable (or defined) in the generic context
 - Not a good surprise to get after deployment!
- Do not rely on abstract performance metrics for model selection
 - Choose and inspect domain-specific performance metrics
 - Compare performance between **sub-groups of interest**
- Consider selective use of models that work well for each sub-group



Thank you!

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