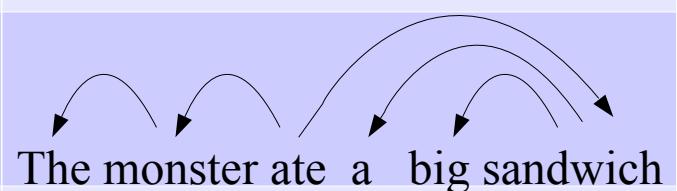
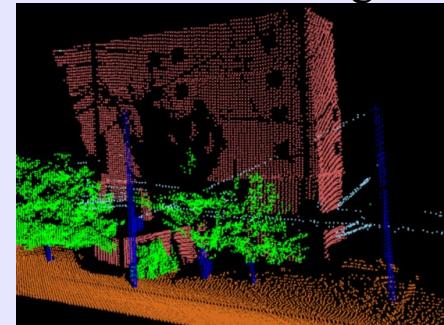


What is structured prediction?

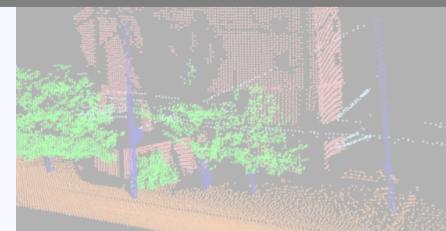
Task	Input	Output
Machine Translation	Ces deux principes se tiennent à la croisée de la philosophie, de la politique, de l'économie, de la sociologie et du droit.	Both principles lie at the crossroads of philosophy, politics, economics, sociology, and law.
Sequence Labeling	The monster ate a big sandwich	Det Noun VerbDetAdj Noun The monster ate a big sandwich
Syntactic Analysis	The monster ate a big sandwich	 The monster ate a big sandwich
3d point cloud classification	3d range scan data	
...many more...		

What is structured prediction?

Task	Input	Output
Machine Tran	Ces deux principes se tiennent à	Both principles lie at the
Se La		
Sy		
Ar		
3d		
classification		
...many more...		

Structured Prediction Haiku

A joint prediction
Across a single input
Loss measured jointly



We want to minimize...

- **Programming complexity.** Most structured predictions are *not* addressed using structured learning because of programming complexity.
- **Test loss.** If it doesn't work, game over.
- **Training time.** Debug/develop productivity, hyperparameter tuning, maximum data input.
- **Test time.** Application efficiency.

Programming complexity

A screenshot of a terminal window titled "search_sequencetask.cc". The window contains C++ code for a class named SequenceTask. The code includes methods for initializing a search object and running it on a vector of examples. The code uses various namespaces like Search, po, and MULTICLASS. The terminal window has a dark background with syntax highlighting for different programming elements. The status bar at the bottom shows the file name, line number (6%), and other system information.

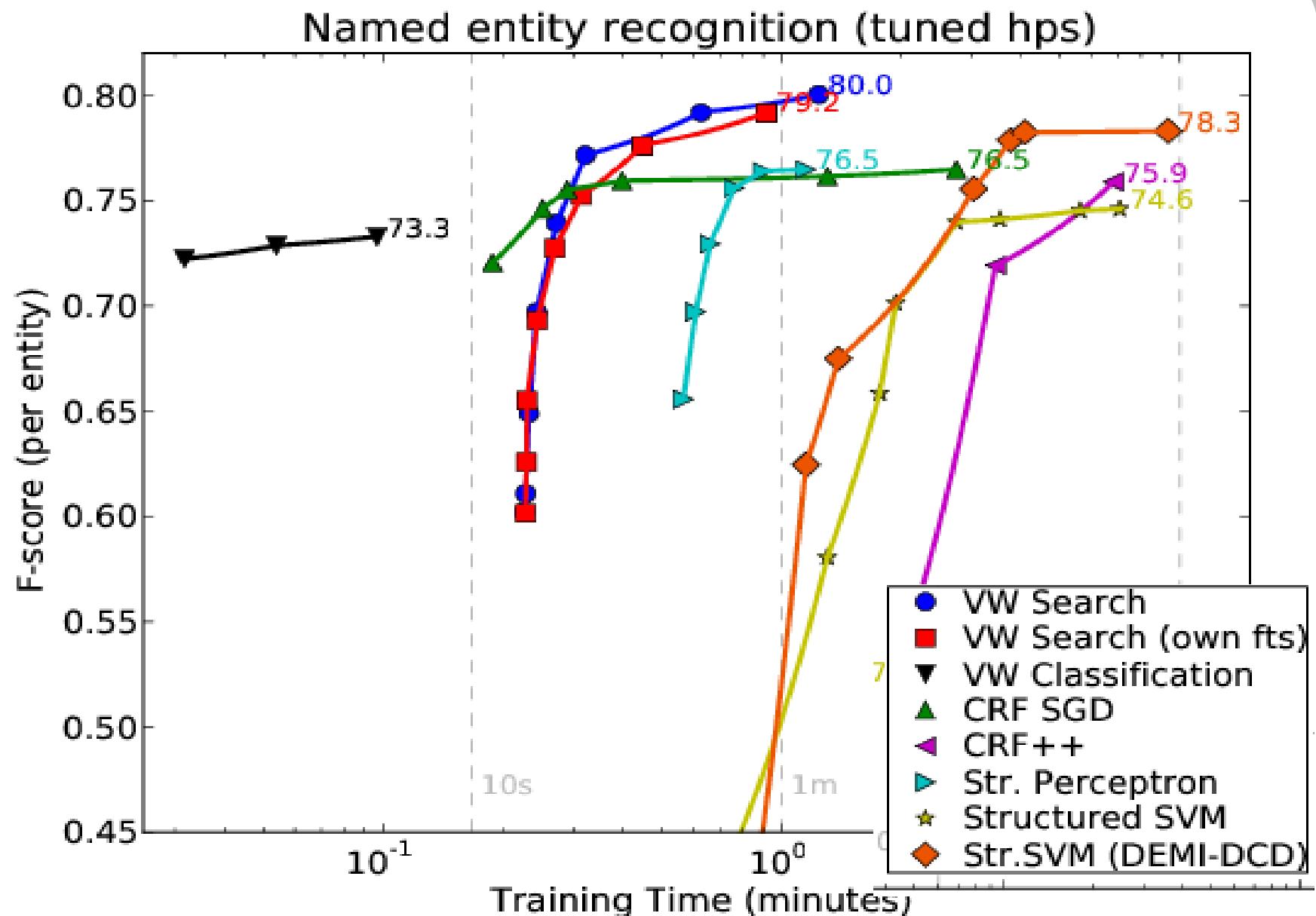
```
search_sequencetask.cc
File Edit Options Buffers Tools C++ YASnippet Development Cscope Help
namespace SequenceTask {
    void initialize(Search::search& sch, size_t& num_actions, po::variables_map& vm) {
        sch.set_options( Search::AUTO_CONDITION_FEATURES |
                         Search::AUTO_HAMMING_LOSS |
                         Search::EXAMPLES_DONT_CHANGE
                         0);
    }

    void run(Search::search& sch, vector<example*>& ec) {
        for (int i=0; i<ec.size(); i++) {
            action oracle      = MULTICLASS::get_example_label(ec[i]);
            size_t prediction = Search::predictor(sch, i+1).set_input(*ec[i]).set_oracle(oracle)
                .set_condition_range(i, sch.get_history_length(), 'p').predict();

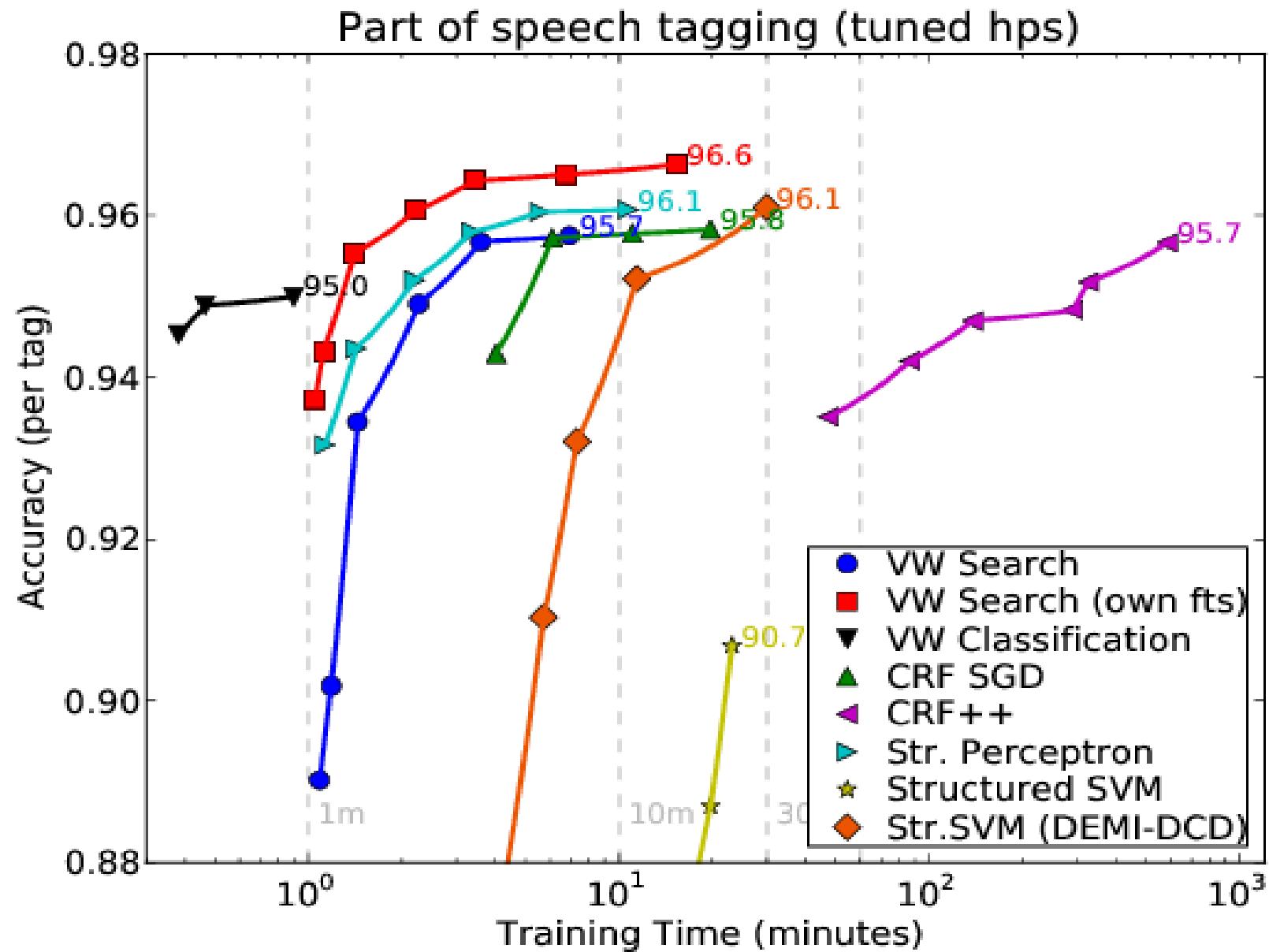
            if (sch.output().good())
                sch.output() << prediction << ' ';
        }
    }
}

-***- search_sequencetask.cc    6% (34,0)    Git-master (C++/1 BufFace AC yas Abbrev)
```

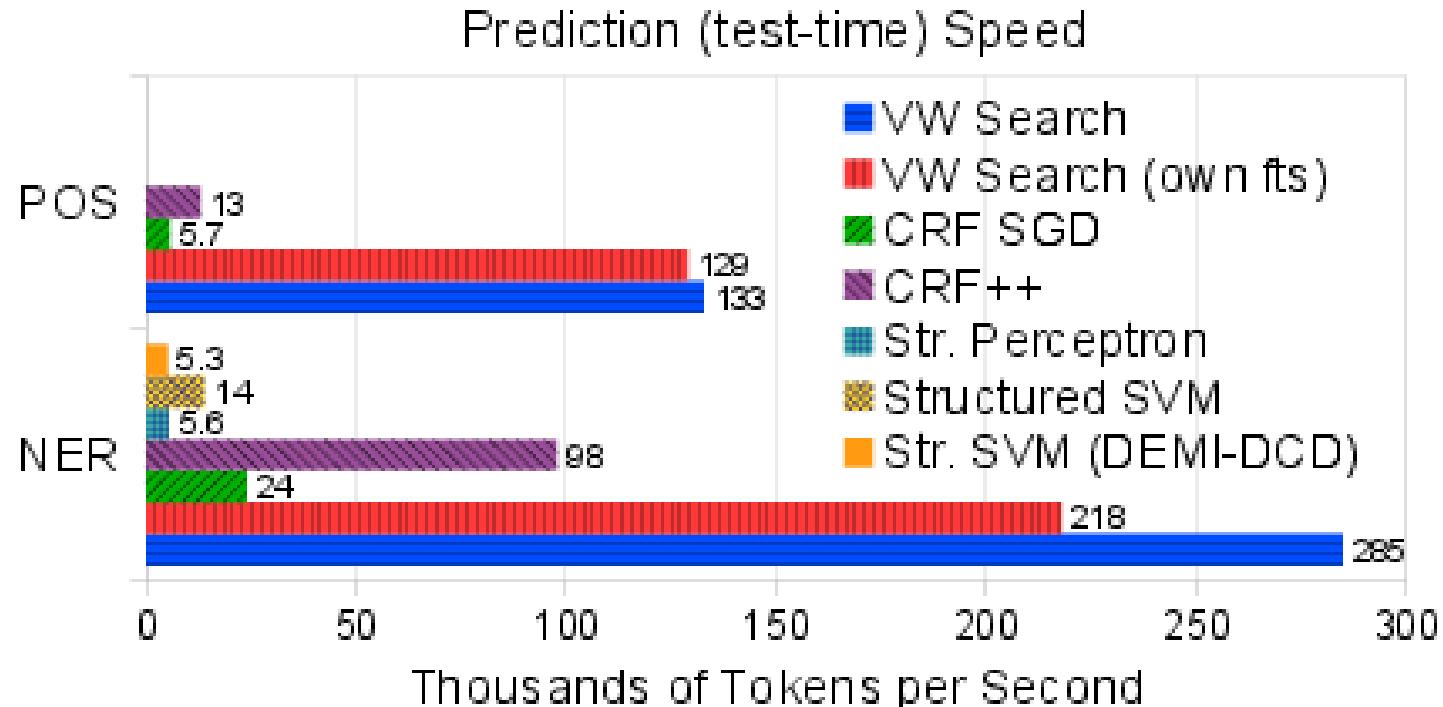
Training time versus test accuracy



Training time versus test accuracy



Test time speed



State of the art accuracy in....

➤ Part of speech tagging (1 million words)

- **vw:** 6 lines of code 3 minutes to train
- **CRFsgd:** 1068 lines 6 minutes
- **CRF++:** 777 lines hours

➤ Named entity recognition (200 thousand words)

- **vw:** 30 lines of code 20 seconds to train
- **CRFsgd:** 1 minute (subopt accuracy)
- **CRF++:** 10 minutes (subopt accuracy)
- **SVMstr:** 876 lines 30 minutes (subopt accuracy)

State of the art accuracy in....

➤ Part of speech tagging (1 million words)

➤ wc:		3.2 seconds
➤ vw:	6 lines of code	3 minutes to train
➤ CRFsgd:	1068 lines	6 minutes
➤ CRF++:	777 lines	hours

➤ Named entity recognition (200 thousand words)

➤ wc:		0.8 seconds
➤ vw:	30 lines of code	20 seconds to train
➤ CRFsgd:		1 minute (subopt accuracy)
➤ CRF++:		10 minutes (subopt accuracy)
➤ SVMstr:	876 lines	30 minutes (subopt accuracy)

Command-line usage

```
% wget http://bilbo.cs.uiuc.edu/~kchang10/tmp/wsj.vw.zip
```

```
% unzip wsj.vw.zip
```

```
% vw -b 24 -d wsj.train.vw -c --search_task sequence \  
--search 45 --search_neighbor_features -1:w,1:w \  
--affix -1w,+1w -f wsj.weights
```

<chat with your neighbor for 3 minutes>

```
% vw -t -i wsj.weights wsj.test.vw
```

<wait 0.15 seconds for 96.4% accuracy>

Python interface to VW

Library interface to VW (*not* a command line wrapper)

It is *actually* documented!!!

Allows you to write code like:

```
import pyvw
vw = pyvw.vw("-quiet")
ex1 = vw.example("1 |x a b |y c")
ex2 = vw.example({'x': ['a', ('b', 1.0)], \
                  'y': ['c']})
ex1.learn()
print ex2.predict()
```

iPython Notebook for Learning to Search

IP[y]: Notebook

Learning_to_Search Last Checkpoint: Oct 03 14:43 (autosaved)

File Edit View Insert Cell Kernel Help



The `_run` function executes the sequence of decisions on a given input. The input will be of whatever type our data is (so, in the above example, it will be a list of (label,word) pairs).

Here is a basic implementation of sequence labeling:

In [39]:

```
class SequenceLabeler(pyvw.SearchTask):
    def __init__(self, vw, sch, num_actions):
        pyvw.SearchTask.__init__(self, vw, sch, num_actions)
        sch.set_options(sch.AUTO_HAMMING_LOSS | sch.AUTO_CONDITION_FEATURES)

    def _run(self, sentence):
        output = []
        for n in range(len(sentence)):
            pos, word = sentence[n]
            with self.vw.example({'w': [word]}) as ex:
                pred = self.sch.predict(examples=ex, my_tag=n+1, oracle=pos, condition=(n, 'p'))
            output.append(pred)
        return output
```

<http://tinyurl.com/pyvwsearch>

<http://tinyurl.com/pyvwtalk>