

Web Mining and Recommender Systems

Tools and techniques for data
processing and visualization

Some helpful ideas for Assignment 2...

1. How can we **crawl our own datasets** from the web?
2. How can we **process those datasets** into structured objects?
3. How can we **visualize and plot** data that we have collected?
4. What libraries can help us to **fit complex models** to those datasets?

Some helpful ideas for Assignment 2...

1. How can we **crawl our own datasets** from the web? → Python requests library + BeautifulSoup
2. How can we **process those datasets** into structured objects? → A few library functions to deal with time+date
3. How can we **visualize and plot** data that we have collected? → Matplotlib
4. What libraries can help us to **fit complex models** to those datasets? → Tensorflow

Web Mining and Recommender Systems

Collecting and parsing Web data with
urllib and BeautifulSoup

Collecting our own datasets

Suppose that we wanted to collect data from a website, but didn't yet have CSV or JSON formatted data

- How could we collect new datasets in machine-readable format?
 - What Python libraries could we use to collect data from webpages?
 - Once we'd collected (e.g.) raw html data, how could we extract structured information from it?

Collecting our own datasets

E.g. suppose we wanted to collect reviews of "The Great Gatsby" from goodreads.com:

(https://www.goodreads.com/book/show/4671.The_Great_Gatsby)



The Great Gatsby
by F. Scott Fitzgerald, Jake Gyllenhaal (Narrator)

★★★★☆ 3.9 ·  Rating details · 3,090,834 Ratings · 56,204 Reviews

A true classic of twentieth-century literature, this edition has been updated by Fitzgerald scholar James L.W. West III to include the author's final revisions and features a note on the composition and text, a personal foreword by Fitzgerald's granddaughter, Eleanor Lanahan—and a new introduction by two-time National Book Award winner Jesmyn Ward.

Want to Read 

Rate this book
★★★★☆

Open Preview

GET A COPY

Kindle Store \$12.99 Amazon Stores ▼ Libraries

Paperback, US / CAN, 180 pages
Published September 2004 by Scribner (first published April 1925)
[More Details...](#) [edit details](#)

Collecting our own datasets

 Nataliya rated it ★★★★★ May 02, 2010
Shelves: childhood-revisited, 2013-reads, i-also-saw-the-film, books-from-childhood-revisited


Oh Gatsby, you old sport, you poor semi-delusionally hopeful dreamer with 'some heightened sensitivity to the promises of life', focusing your whole self and soul on that elusive money-colored green light – a dream that shatters just when you are *this* close to it.



Jay Gatsby, who dreamed a dream with the passion and courage few possess – and the tragedy was that it was a wrong dream colliding with reality that was even more wrong – and deadly.

Just like the Great Houdini – the association the [...more](#)

713 likes · Like · see review

 Alex rated it ★★★★★ Dec 24, 2007
The Great Gatsby is your neighbor you're best friends with until you find out he's a drug dealer. It charms you with some of the most elegant English prose ever published, making it difficult to discuss the novel without the urge to stammer awestruck about its beauty. It would be evidence enough to

How could we extract fields including

- The *ID* of the user,
- The *date* of the review
- The *star rating*
- The *text* of the review itself?
- The *shelves* the book belongs to

Reading the html data

This isn't very nice to look at, it can be easier to read in a browser or a text editor (which preserves formatting):

```
1 <!DOCTYPE html>
2 <html class="desktop
3 ">
4
5
6 <head prefix="og: http://ogp.me/ns# fb: http://ogp.me/ns/fb# good_reads: http://ogp.me/ns/fb/good_reads#">
7   <title>
8     The Great Gatsby by F. Scott Fitzgerald
9   </title>
10
11
12   <script type="text/javascript"> var ue_t0=window.ue_t0||+new Date();
13 </script>
14   <script type="text/javascript">
15     (function(e){var c=e;var a=c.ue||{};a.main_scope="mainscopeesm";a.q=[];a.t0=c.ue_t0||+new Date();a.d=g;fu
{m:m,f:l,l:h,c:""+j,err:i,fromOnError:1,args:arguments};c.ueLogError(k);return false}b.skipTrace=1;e.onerror=
{e.attachEvent("onload",f)}}a.tag=d("tag");a.log=d("log");a.reset=d("rst");c.ue_csm=c;c.ue=a;c.ueLogError=d("
f=d.head||d.getElementsByTagName("head")[0]||d.documentElement,h=d.createElement("script");h.async="async";h.
amazon.com",j="/images/G/01/csminstrumentation/",h=e.ue_file||"ue-full-11e51f253e8ad9d145f4ed644b40f692._V1_
1:0}i=f?"https://":"http://";i+=f?g:k;i+=j;i+=h;c(i)}if(!e.ue_inline){if(a.loadUEFull){a.loadUEFull()}else{b(
```

Reading the html data

To extract review data, we'll need to look for the part of the html code which contains the reviews:

```
1227 <div id="bookReviews">
```

Here it is (over 1000 lines into the page!)

```
1233 <div class="friendReviews elementListBrown" >
1234 <div class="section firstReview">
```

```
1237 <div id="review_101057684" class="review nosyndicate" itemprop="reviews" itemscope itemType="http://schema.org/Review">
1238 <link itemprop="url" href="https://www.goodreads.com/review/show/101057684" />
1239 <a title="Nataliya" class="left imgcol" href="/user/show/3672777-nataliya">
```

```
1241 <div class="left bodycol">
```

```
1242 <div class="reviewHeader uitext stacked">
```

```
1243 <a class="reviewDate createdAt right" href="/review/show/101057684?book_show_action=true">May 02, 2010</a>
```

```
1245 <span itemprop="author" itemscope itemType="http://schema.org/Person">
```

```
1246 <a title="Nataliya" class="user" itemprop="url" name="Nataliya" href="/user/show/3672777-nataliya">Nataliya</a>
```

```
1247 </span>
```

```
1248
```

Reading the html data

To extract review data, we'll need to look for the part of the html code which contains the reviews:

```
<div id="review_101057684" class="review nosyndicate" id="review_101057684">
  <link itemprop="url" href="https://www.goodreads.com/review/show/101057684">
  <a title="Nataliya" class="left imgcol" href="/user/show/Nataliya">
```

- Note that each individual review starts with a block containing the text "<div id="review_..."
- We can collect all reviews by looking for instances of this text

Code: parsing the review contents

Next we have to write a method to parse individual reviews (i.e., given the text of one review, extract formatted fields into a dictionary)

```
In [8]: def parseReview(review):
        d = {}
        d['stars'] = review.split('<span class="staticStars" title="')[1].split('')[0]
        d['date'] = review.split('<a class="reviewDate')[1].split('>')[1].split('<')[0]
        d['user'] = review.split('<a title="')[1].split('')[0]
        shelves = []
        try:
            shelfBlock = review.split('<div class="uitext greyText bookshelves">')[1].split('</div')[0]
            for s in shelfBlock.split('shelf=')[1:]:
                shelves.append(s.split('')[0])
            d['shelves'] = shelves
        except Exception as e:
            pass
        reviewBlock = review.split('<div class="reviewText stacked">')[1].split('</div')[0]
        d['reviewBlock'] = reviewBlock
        return d
```

Code: parsing the review contents

Let's look at it line-by-line:

```
In [8]: def parseReview(review):  
        d = {}
```

- We start by building an empty dictionary
- We'll use this to build a *structured* version of the review

Code: parsing the review contents

Let's look at it line-by-line:

Note: Two splits: everything *after* the first quote, and *before* the second quote

- The next line is more complex:

```
d['stars'] = review.split('<span class=" staticStars" title="')[1].split('')[0]
```

- We made this line by noticing that the stars appear in the html inside a span with class " staticStars":

```
1248     rated it
1249     <span class=" staticStars" title="it was amazing"><span size="15x15" class="staticStar p10">it was amazing<
1250 p10"></span><span size="15x15" class="staticStar p10"></span></span>
```

- Our "split" command then extracts everything inside the "title" quotes

Code: parsing the review contents

Let's look at it line-by-line:

- The following two lines operate in the same way:

Note: Everything between the two brackets of this "<a" element

```
d['date'] = review.split('<a class="reviewDate"')[1].split('>')[1].split('<')[0]
d['user'] = review.split('<a title="')[1].split('"')[0]
```

- Again we did this by noting that the "date" and "user" fields appear inside certain html elements:

```
1240
1241 <div class="left bodycol">
1242   <div class="reviewHeader uitrust stacked">
1243     <a class="reviewDate createdAt right" href="/review/show/101057684?book_show_action=true">May 02, 2010</a>
1244
1245     <span itemprop="author" itemscope itemtype="http://schema.org/Person">
1246       <a title="Nataliya" class="user" itemprop="url" name="Nataliya" href="/user/show/3672777-nataliya">Nataliya</a>
1247     </span>
1248
```


Code: parsing the review contents

Let's look at it line-by-line:

- Next we extract the "shelves" the book belongs to
- This follows the same idea, but in a "for" loop since there can be many shelves per book:

```
shelves = []
try:
    shelfBlock = review.split('<div class="uitext greyText bookshelves">')[1].split('</div>')[0]
    for s in shelfBlock.split('shelf=')[1:]:
        shelves.append(s.split(' ')[0])
    d['shelves'] = shelves
except Exception as e:
    pass
```

Note: Everything inside a particular <div

- Here we use a try/except block since this text will be missing for users who didn't add the book to any shelves

Code: parsing the review contents

Next let's extract the review contents:

```
In [8]: def parseReview(review):
        d = {}
        d['stars'] = review.split('<span class=" staticStars" title="')[1].split('')[0]
        d['date'] = review.split('<a class="reviewDate')[1].split('>')[1].split('<')[0]
        d['user'] = review.split('<a title="')[1].split('')[0]
        shelves = []
        try:
            shelfBlock = review.split('<div class="uitext greyText bookshelves">')[1].split('</div')[0]
            for s in shelfBlock.split('shelf=')[1:]:
                shelves.append(s.split('')[0])
            d['shelves'] = shelves
        except Exception as e:
            pass
        reviewBlock = review.split('<div class="reviewText stacked">')[1].split('</div')[0]
        d['reviewBlock'] = reviewBlock
        return d
```

Code: parsing the review contents

Now let's look at the results:

```
In [9]: reviewDict = [parseReview(r) for r in reviews]
```

```
In [10]: reviewDict[0]
```

```
Out[10]: {'date': 'May 02, 2010',  
  'reviewBlock': '\\n  
    <span id="reviewTextContainer101057684" class="readable"\\n  
    >\\n  
    \\n<span id="freeTextContainer7513160808421149349"><br>Oh Gatsby, you old sport, you poor semi-delusionally h  
    i>\\n', focusing your whole self and  
    you are *this* close to it. <br><br>  
    dimages/1380334543i/693798_SX540_.j  
    eam with the passion and courage few  
    hat was even more wrong - and deadl  
    n id="freeText7513160808421149349" s  
    eful dreamer with \\n'<i>some height  
    bul on that elusive money-colored gr  
     <br><br>Just like the Great Ho  
    ed in illusions and escape. Except e  
    be the world, our past, and ourselve  
    s, giving rise to one of the most famous closing lines of a novel.<blockquote>\\n <i>\\n'Gatsby believed in the gre  
    en light, the orgastic future that year by year recedes before us. It eluded us then, but that\\xe2\\x80\\x99s no ma  
    tter \\xe2\\x80\\x99s to me, for as long as I can reach out my arms, I can touch the Great Gatsby. And one fine morning \\xe2\\x80\\x99s
```

- Looks okay, but the review block itself still contains embedded html (e.g. images etc.)
- How can we extract just the text part of the review?

The BeautifulSoup library

Extracting the text contents from the html review block would be extremely difficult, as we'd essentially have to write a html parser to capture all of the edge cases

Instead, we can use an existing library to parse the html contents: **BeautifulSoup**

Code: parsing with BeautifulSoup

BeautifulSoup will build an element tree from the html passed to it. For the moment, we'll just use it to extract the text from a html block

```
In [11]: from bs4 import BeautifulSoup
```

```
In [12]: soup = BeautifulSoup(reviewDict[0]['reviewBlock'])
```

```
In [13]: soup.text
```

```
Out[13]: "\\n          \\n          \\nOh Gatsby, you old sport, you poor semi-delusionally hopeful dreamer with \\\"some heightened sensitivity to the promises of life\\\", focusing your whole self and soul on that elusive money-colored green light - a dream that shatters just when you are *this* close to it. Jay Gatsby, who dreamed a dream with the passion and courage few possess - and the tragedy was that it was a wrong dream colliding with reality that was even more wrong - and deadly. Just like the Great Houdini - the association the\\n Oh Gatsby, you old sport, you poor semi-delusionally hopeful dreamer with \\\"some heightened sensitivity to the promises of life\\\", focusing your whole self and soul on that elusive money-colored green light - a dream that shatters just when you are *this* close to it. Jay Gatsby, who dreamed a dream with the passion and courage few possess - and the tragedy was that it was a wrong dream colliding with reality that was even more wrong - and deadly. Just like the Great Houdini - the association the title of this book so easily invokes - you specialized in illusions and escape. Except even the power of most courageous dreamers can be quite helpless to allow us escape the world, our past, and ourselves, giving rise to one of
```

The BeautifulSoup library

In principle we could have used BeautifulSoup to extract *all* of the elements from the webpage

However, for simple page structures, navigating the html elements is not (necessarily) easier than using primitive string operations

Advanced concepts...

1. What if we have a webpage that loads content **dynamically**?

(e.g. https://www.amazon.com/gp/profile/amzn1.account.AHQSDGUKX6BESSVAOWMIAJKBOZPA/ref=cm_cr_dp_d_gw_tr?ie=UTF8)

- The page (probably) uses javascript to generate requests for new content
- By monitoring network traffic, perhaps we can view and reproduce those requests
- This can be done (e.g.) by using the Developer Tools in chrome

Pages that load dynamically...

The screenshot shows an Amazon profile page in a browser. The address bar displays the URL: `https://www.amazon.com/gp/profile/amzn1.account.AHQSDGUKX6BE5SVAOWMIAJKBOZPA/ref=cm_cr_dp_d_gw_tr?ie=UTF8`. The browser's developer tools are open to the Network tab, showing a list of requests. A green arrow points from the text "Scroll to bottom..." to the "Order" link in the navigation bar.

Scroll to bottom...

Name	Status	Type	Initiator	Size	Time	Waterfall
<input type="checkbox"/> OE/	204	text/...	ClientSide...	164 B	79 ms	
<input type="checkbox"/> OE/	204	text/...	ClientSide...	164 B	76 ms	

2 requests | 328 B transferred

Console | What's New | Request blocking

Highlights from the Chrome 70 update

Live Expressions in the Console
Pin expressions to the top of the Console to monitor their values in real-time.

Highlight DOM nodes during Eager Evaluation
Type an expression that evaluates to a node to highlight that node in the viewport.

Autocomplete Conditional Breakpoints

Pages that load dynamically...

The screenshot shows a web browser displaying an Amazon product page. The browser's developer tools are open, showing the Network tab. A callout box with the text "Look at requests that get generated" points to a specific XHR request in the network log. The request is a POST to a URL ending in "visitor?nextPageToken=...". The network log shows a list of requests with columns for Name, Status, Type, Initiator, Size, Time, and Waterfall. The highlighted request is a 5.7 KB XHR request that took 210 ms to complete. Below the network log, the console shows "31 requests | 20.8 KB transferred".

Look at requests that get generated

Name	Status	Type	Initiator	Size	Time	Waterfall
trackNextPageToken	200	xhr	customer-p...	0 B	21 ms	
visitor?nextPageToken=...	200	xhr	customer-p...	5.7 KB	210 ms	
81YvKtMEL.jpg	200	jpeg	customer-p...	(fro...	23 ms	
61YtqXulrBL.jpg	200	jpeg	customer-p...	(fro...	14 ms	
61PPEQH5xL.jpg	200	jpeg	customer-p...	(fro...	15 ms	
81xmeCLsYL.jpg	200	jpeg	customer-p...	(fro...	23 ms	
81eLwJrHrFL.jpg	200	jpeg	customer-p...	(fro...	22 ms	
918x3jffuOL.jpg	200	jpeg	customer-p...	(fro...	47 ms	
61d9iHybt8L.jpg	200	jpeg	customer-p...	(fro...	17 ms	
21ID+J39k3L.jpg	200	jpeg	customer-p...	(fro...	15 ms	
81XoNLkc61L.jpg	200	jpeg	customer-p...	(fro...	20 ms	
OE/	204	text/...	ClientSide...	164 B	77 ms	
OE/	204	text/...	ClientSide...	165 B	80 ms	
OE/	204	text/...	ClientSide...	164 B	78 ms	
OE/	204	text/...	ClientSide...	165 B	79 ms	
OE/	204	text/...	ClientSide...	165 B	75 ms	
OE/	204	text/...	ClientSide...	164 B	79 ms	
OE/	204	text/...	ClientSide...	164 B	83 ms	
OE/	204	text/...	ClientSide...	165 B	84 ms	

31 requests | 20.8 KB transferred

Pages that load dynamically...

Let's try to reproduce this request

Profile
https://www.amazon.com/gp/profile/amzn1.account.AHQSDGUKX6BESSVAOWMIAJKBOZPA/ref=cm_cr_dp_d_gw_tr?ie=UTF8

See full review

Accord APSWDF Adjustable Magnetic Air Deflector for Sidewall and Ceiling Registers
★★★★☆ 527

Amazon Customer reviewed a product · Oct 12, 2016

★★★★☆ Verified Purchase
Works but tastes awful. Hard to master.
Can't imagine the chemicals in this are very safe. It tastes like acetone polish

Amazon Customer reviewed a product · Oct 12, 2016

★★★★☆ Verified Purchase
THE HORRIBLE SMELL!!!
I have had this feeder for just a couple of weeks. I feed my cat fancy feast cans twice a day in my kitchen. At the end of the day I remove the bowl and lid, rinse out any old food, and hand wash with a little dish soap....

See full review
20 helpful votes

Network

Filter: All XHR JS CSS Img Media Font Doc WS Manifest Other

dynamic-menu.html?wishlist...
sidepanel?rid=FJ1V86RJFVP...
notification?communicationC...
71jCV74H3cL_AC_UL75_SR7...
51O%2BFT4nOpL_AC_UL75...
61487dKasVL_AC_UL75_Sk7...
track?category=Profile&even...
visitor?nextPageToken=%7B...
81jvPEthHEL.jpg
61YtqXulrBL.jpg
61PFEQH5xiL.jpg
81xmeCLsyll.jpg
81eLWJrHrFL.jpg
918x3jffuOL.jpg
61d9iHybt8L.jpg
21ID+J39k3L.jpg
81XoNLkc61L.jpg
OE/
OE/
OE/

40 requests | 22.2 KB transferred

Headers

General

Request URL: https://www.amazon.com/profilewidget/timezone/visitor?nextPageToken=%7B%22st%22%3A%7B%22n%22%3A%22148626335100%22%7D%2C%22ctrbnId%22%3A%7B%22%22%3A%22amzn1.productreview.R1TH4IN%22%3A%7B%22%22%3A%7B%22%22%3A%22amzn1.ac...
account.AHQSDGUKX6BESSVAOWMIAJKBOZPA.customer.ATVPDKIX0DER%22%7D%2C%22filteredContributionTypes=productreview%2Cglimpse%2Cideasfor...
ecteur.amzn1.account.AHQSDGUKX6BESSVAOWMIAJKBOZPA

Request Method: GET
Status Code: 200
Remote Address: 13.35.96.165:443
Referrer Policy: no-referrer-when-downgrade

Response Headers

content-encoding: gzip
content-type: application/json;charset=UTF-8
date: Wed, 07 Nov 2018 00:21:17 GMT
server: Server
set-cookie: session-id=140-8603009-6912711; Domain=.amazon.com; Expires=Tue, 01-Jan-2036 08:00:01 GMT; Path=/

Console

What's New

Request blocking

Highlights from the Chrome 70 update

Live Expressions in the Console
Pin expressions to the top of the Console to monitor their values in real-time.

Highlight DOM nodes during Eager Evaluation
Type an expression that evaluates to a node to highlight that node in the viewport.

Autocomplete Conditional Breakpoints

Advanced concepts...

2. What if we require **passwords, captchas, or cookies?**

- You'll probably need to load an actual browser
- This can be done using a **headless browser**, i.e., a browser that is controlled via Python
 - I usually use **splinter** (<https://splinter.readthedocs.io/en/latest/>)
- Note that once you've entered the password, solved the captcha, or obtained the cookies, you can normally continue crawling using the *requests* library

Summary

- Introduced programmatic approaches to collect datasets from the web
- The **urllib** library can be used to request data from the web as if it is a file, whereas **BeautifulSoup** can be used to convert the data to structured objects
 - Parsing can also be achieved using primitive string processing routines
- Make sure to check the page's **terms of service** first!

Web Mining and Recommender Systems

Parsing time and date data

Time and date data

Dealing with time and date data can be difficult as string-formatted data doesn't admit easy comparison or feature representation:

- Which date occurs first, 4/7/2003 or 3/8/2003?
- How many days between 4/5/2003 - 7/15/2018?
- e.g. how many hours between 2/6/2013 23:02:38 - 2/7/2013 08:32:35?

Time and date data

Most of the data we've seen so far include plain-text time data, that we need to carefully manipulate:

```
{'business_id': 'FYWN1wneV18bWNgQjJ2GNg', 'attributes':  
{'BusinessAcceptsCreditCards': True, 'AcceptsInsurance':  
True, 'ByAppointmentOnly': True}, 'longitude': -111.9785992,  
'state': 'AZ', 'address': '4855 E Warner Rd, Ste B9',  
'neighborhood': 'Phoenix', 'city': 'Phoenix', 'hours': {'Tuesday':  
'7:30-17:00', 'Wednesday': '7:30-17:00', 'Thursday': '7:30-  
17:00', 'Friday': '7:30-17:00', 'Monday': '7:30-17:00'},  
'postal_code': '85044', 'review_count': 22, 'stars': 4.0,  
'categories': ['Dentists', 'General Dentistry', 'Health &  
Medical', 'Oral Surgeons', 'Cosmetic Dentists',  
'Orthodontists'], 'is_open': 1, 'name': 'Dental by Design',  
'latitude': 33.3306902}
```

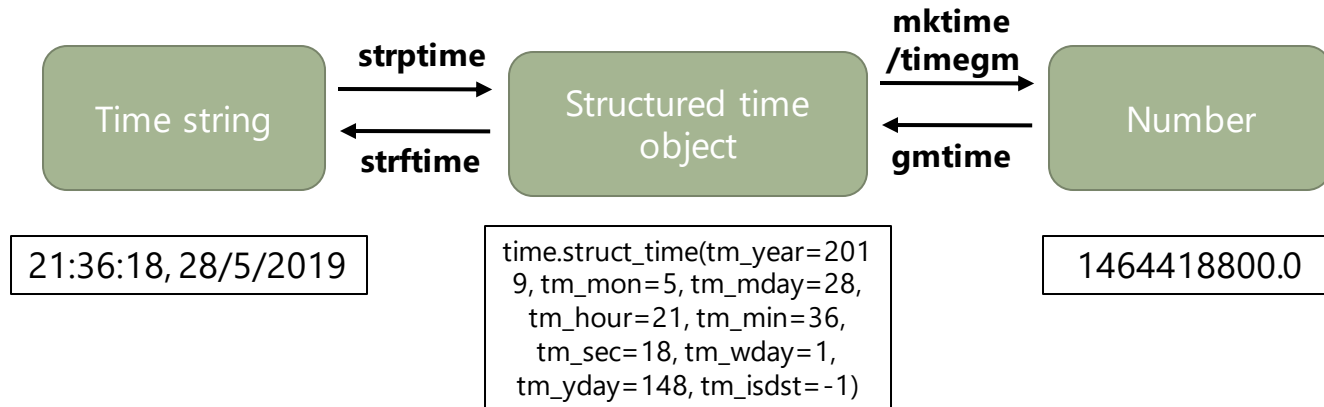

Time and date data

Here we'll cover a few functions:

- Time.strptime: convert a time **string** to a structured time **object**
- Time.strftime: convert a time **object** to a **string**
- Time.mktime / calendar.timegm: convert a time **object** to a **number**
- Time.gmtime: convert a **number** to a time **object**

Time and date data

Here we'll cover a few functions:



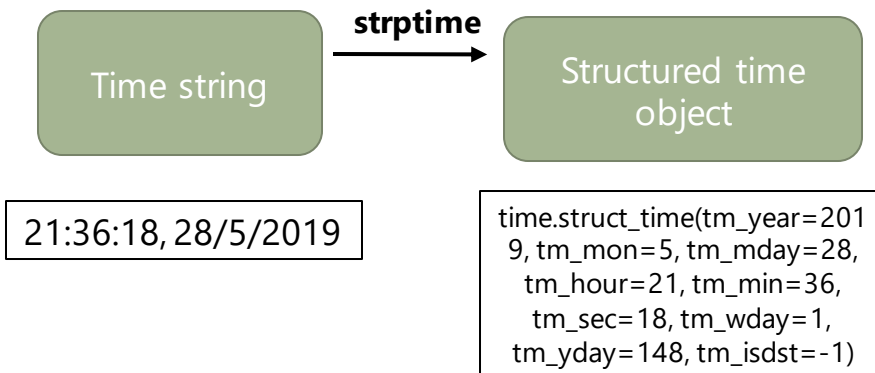
Concept: Unix time

Internally, time is often represented as a number, which allows for easy manipulation and arithmetic

- The value (Unix time) is the **number of seconds since Jan 1, 1970 in the UTC timezone**
- so I made this slide at 1532568962 = 2018-07-26 01:36:02 UTC (or 18:36:02 in my timezone)
- But real datasets generally have time as a "human readable" string
- Our goal here is to convert between these two formats

strptime

First, let's look at converting a string to a structured object (strptime)



Code: time.strptime()

```
In [1]: import time
import calendar
```

String-formatted time data

```
In [2]: timeString = "2018-07-26 01:36:02"
```

```
In [3]: timeStruct = time.strptime(timeString, "%Y-%m-%d %H:%M:%S")
```

```
In [4]: timeStruct
```

```
Out[4]: time.struct_time(tm_year=2018, tm_mon=7, tm_mday=26, tm_hour=1, tm_min=36, tm_sec=2, tm_wday=3, tm_yday=207, tm_isdst=-1)
```

```
In [5]: timeStruct.tm_wday
```

Note: this day is a Wednesday!

```
Out[5]: 3
```

```
In [6]: help(time.strptime)
```

```
Help on built-in function.strptime in module time:
```

```
strptime(...)
strptime(string, format) -> struct_time
```

Note: different time formatting options in the help page

```
Parse a string to a time tuple according to a format specification.
```

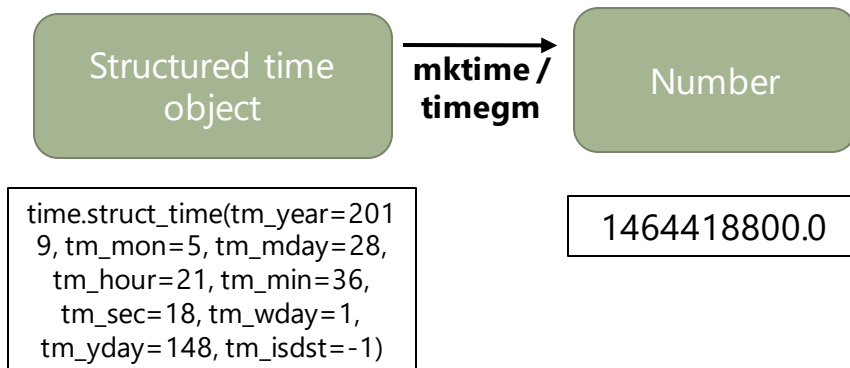
strptime

Strptime is convenient when we want to extract **features** from data


- E.g. does a date correspond to a weekday or a weekend?
- Converting month names or abbreviations (e.g. "Jan") to month numbers
- Dealing with mixed-format data by converting it to a common format
- But if we want to perform arithmetic on timestamps, converting to a number may be easier

time.mktime and calendar.timegm

For this we'll use mktime to convert our structured time object to a number:




Code: time.mktime() and calendar.timegm()

```
In [7]: t1 = calendar.timegm(timeStruct)  Structured time data from previous slide
```

```
In [8]: t2 = time.mktime(timeStruct)
```

```
In [9]: t1, t2
```

```
Out[9]: (1532568962, 1532594162.0)
```

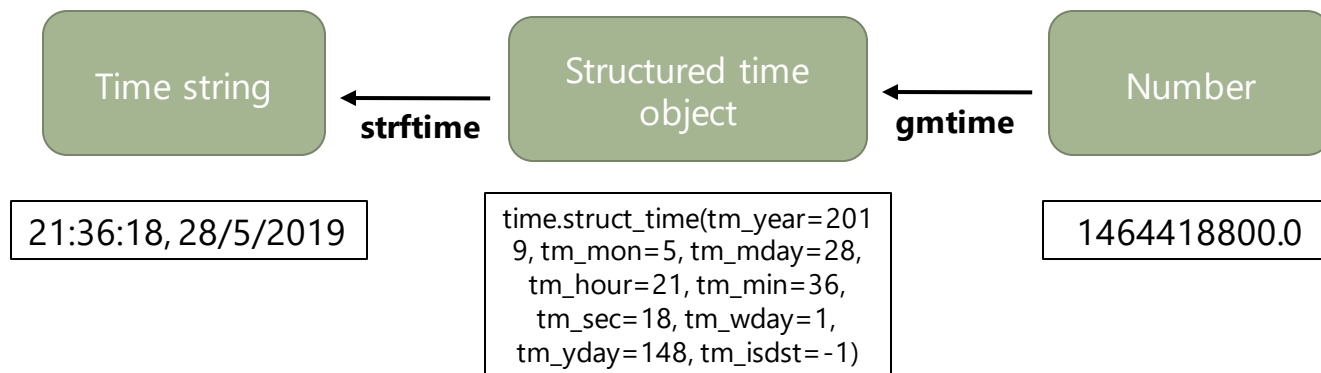
```
In [10]: t1 + 60*60*24*5 
```

```
Out[10]: 1533000962 Five days later
```

- time.mktime() allows us to convert our structured time object to a number
- **NOTE:** mktime assumes the structure is a *local* time whereas timegm assumes the structure is a *UTC* time
- This allows for easy manipulation, arithmetic, and comparison (e.g. sorting) of time data

time.strptime and time.gmtime

Finally, both of these operations can be *reversed*, should we wish to format time data as a string or structure



Code: time.strftime() and time.gmtime()

```
In [11]: time.gmtime(t1 + 60*60*24*5) ← Five days later than the previous time
```

```
Out[11]: time.struct_time(tm_year=2018, tm_mon=7, tm_mday=31, tm_hour=1, tm_min=36, tm_sec=2, tm_wday=1, tm_yday=212, tm_isds  
t=0)
```

```
In [12]: time.strftime("%Y-%m-%d %H:%M:%S", time.gmtime(t1 + 60*60*24*5))
```

```
Out[12]: '2018-07-31 01:36:02'
```

- These methods can be used to put adjusted times back into string format

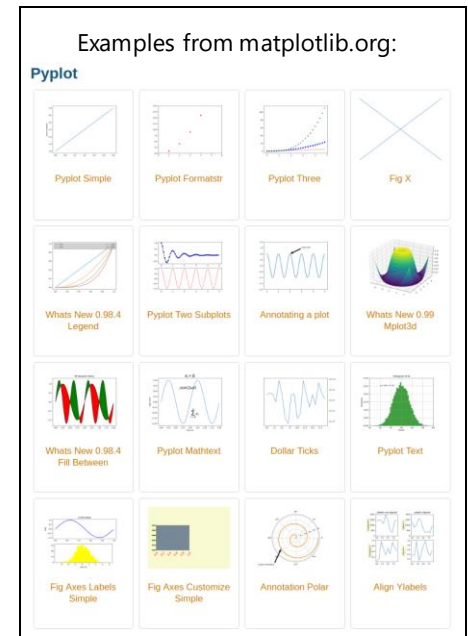
Web Mining and Recommender Systems

Introduction to Matplotlib

Matplotlib

Matplotlib is a powerful library that can be used to generate both quick visualizations, as well as publication-quality graphics

- We'll introduce some of its most basic functionality (via pyplot), such as bar and line plots
- Examples (with code) of the types of plots that can be generated are available on <https://matplotlib.org/>



Code: generating some simple statistics

First, let's quickly compile some statistics from (e.g.) Yelp's review data

```
In [1]: import json
import time
path = "datasets/yelp_data/review.json"
f = open(path, 'r')
```

```
In [2]: dataset = []
for i in range(50000):
    dataset.append(json.loads(f.readline()))
```

```
In [3]: datasetWithTimeValues = []
```

```
In [4]: for d in dataset:
    d['date']
    d['timeStruct'] = time.strptime(d['date'], "%Y-%m-%d")
    d['timeInt'] = time.mktime(d['timeStruct'])
    datasetWithTimeValues.append(d)
```

Code: generating some simple statistics

```
In [5]: from collections import defaultdict
```

```
In [6]: weekRatings = defaultdict(list)
```

```
In [7]: for d in datasetWithTimeValues:  
        day = d['timeStruct'].tm_wday  
        weekRatings[day].append(d['stars'])
```

```
In [8]: weekAverages = {}
```

```
In [9]: for d in weekRatings:  
        weekAverages[d] = sum(weekRatings[d]) * 1.0 / len(weekRatings[d])
```

```
In [10]: weekAverages
```

```
Out[10]: {0: 3.7094594594594597,  
         1: 3.715375187253166,  
         2: 3.750551876379691,  
         3: 3.763665361751486,  
         4: 3.7551891653172382,  
         5: 3.7231843981953134,  
         6: 3.7072147651006713}
```

← Average ratings per day of week

Code: drawing a simple plot

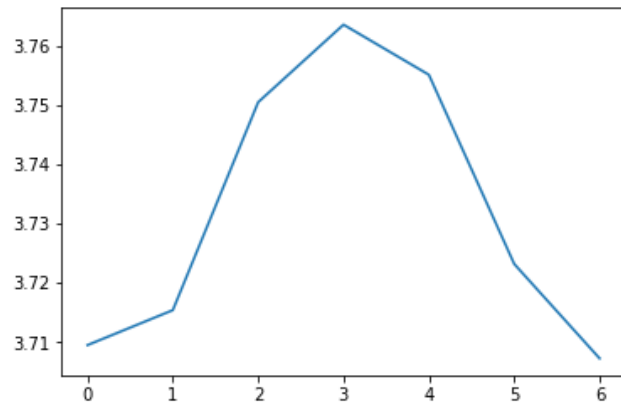
```
In [11]: X = list(weekAverages.keys()) ← [0,1,2,3,4,5,6]
```

```
In [12]: Y = [weekAverages[x] for x in X]
```

```
In [13]: import matplotlib.pyplot as plt
```

```
In [14]: plt.plot(X, Y)
```

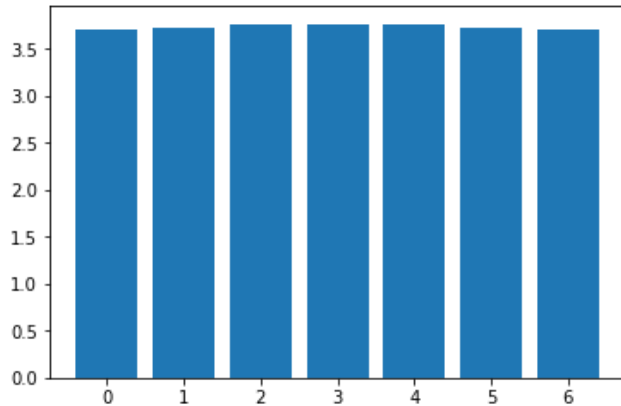
```
Out[14]: [<matplotlib.lines.Line2D at 0x7fc15a615a20>]
```



Code: bar plots

```
In [15]: plt.bar(X, Y)
```

```
Out[15]: <Container object of 7 artists>
```

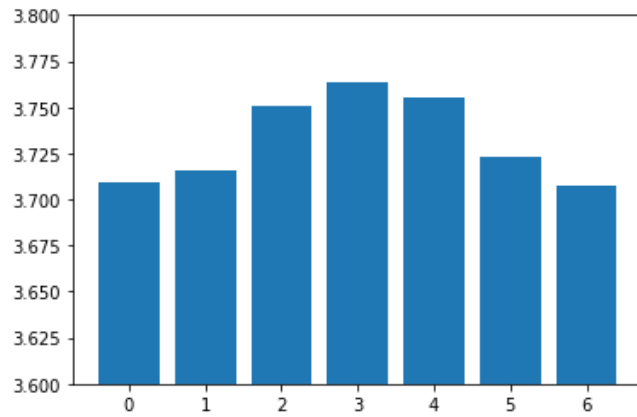


- Looks right, but need to zoom in more to see the detail

Code: bar plots

```
In [16]: plt.ylim(3.6, 3.8)  
plt.bar(X, Y)
```

```
Out[16]: <Container object of 7 artists>
```

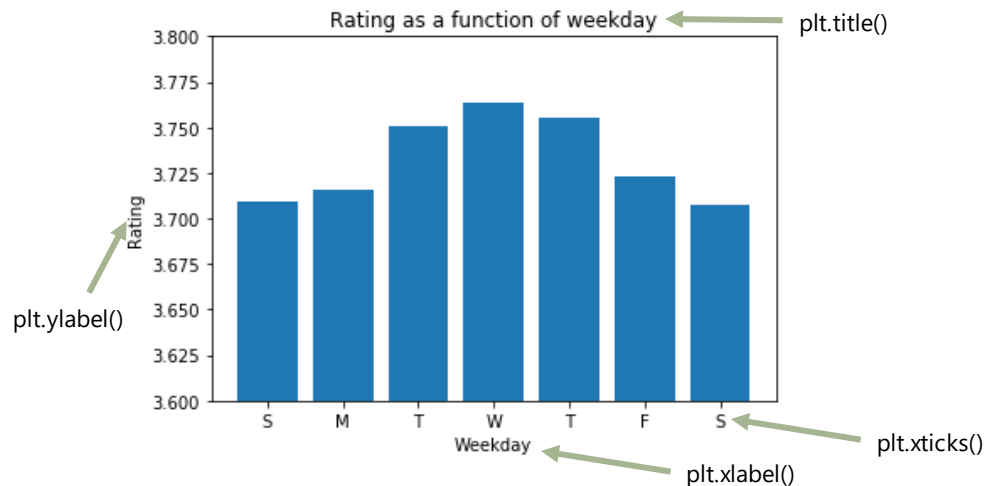


- Next let's add some details

Code: bar plots

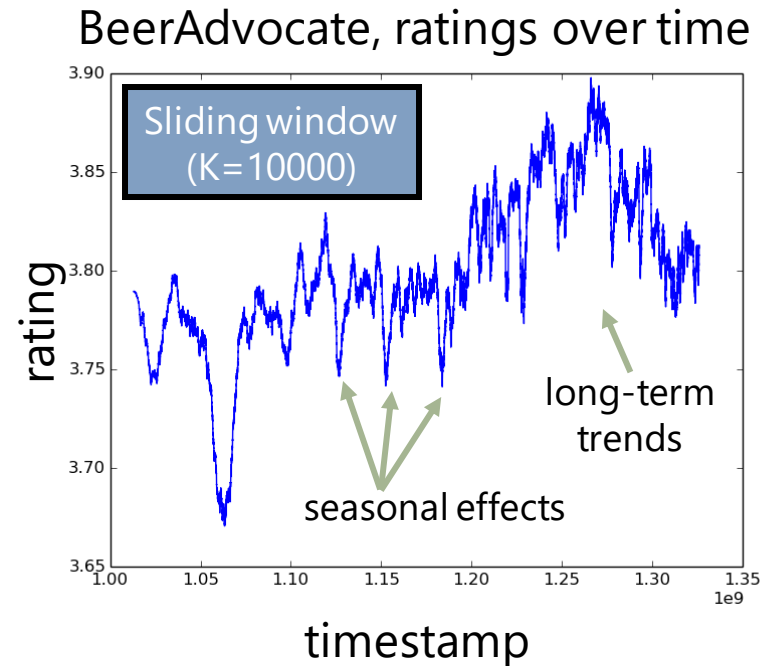
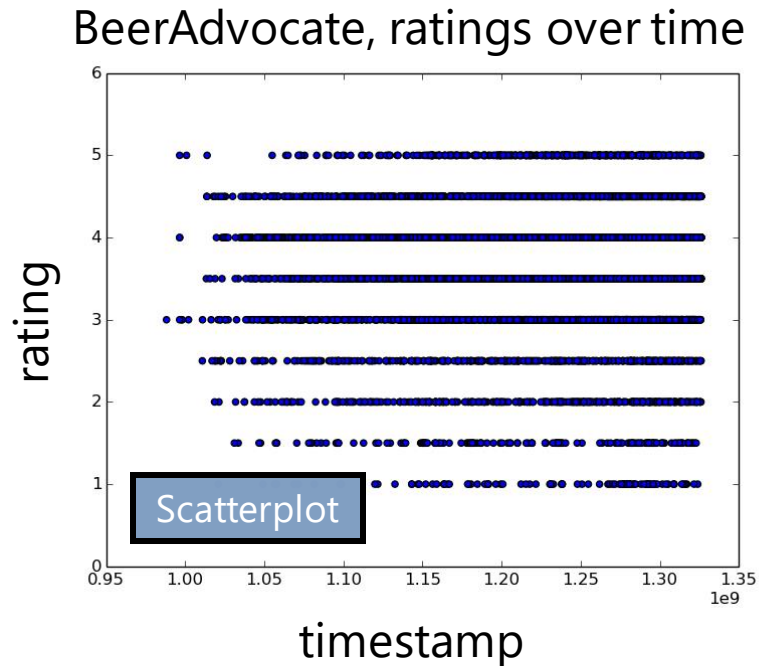
```
In [17]: plt.ylim(3.6, 3.8)
plt.xlabel("Weekday")
plt.ylabel("Rating")
plt.xticks([0,1,2,3,4,5,6],['S', 'M', 'T', 'W', 'T', 'F', 'S'])
plt.title("Rating as a function of weekday")
plt.bar(X, Y)
```

Out[17]: <Container object of 7 artists>



Example: sliding windows

Also useful to plot data:



Code on:

<http://jmcauley.ucsd.edu/code/week10.py>

Web Mining and Recommender Systems

Gradient descent in tensorflow

Tensorflow

Tensorflow, though often associated with deep learning, is really just a library that simplifies gradient descent and optimization problems, like those we've already implemented

Most critically, it computes gradients **symbolically**, so that you can just specify the objective, and Tensorflow can run gradient descent

Here we'll reimplement some of our previous gradient descent code in tensorflow

Code: Gradient Descent in Tensorflow

Reading the data is much the same as before (except that we first import the tensorflow library)

```
In [1]: import tensorflow as tf
```

```
In [2]: path = "datasets/PRSA_data_2010.1.1-2014.12.31.csv"
f = open(path, 'r')
```

```
In [3]: dataset = []
header = f.readline().strip().split(',')
for line in f:
    line = line.split(',')
    dataset.append(line)
```

```
In [4]: header.index('pm2.5')
```

```
Out[4]: 5
```

```
In [5]: dataset = [d for d in dataset if d[5] != 'NA']
```

Code: Gradient Descent in Tensorflow

Next we extract features from the data

```
In [6]: def feature(datum):  
        feat = [1, float(datum[7]), float(datum[8]), float(datum[10])] # Temperature, pressure, and wind speed  
        return feat
```

```
In [7]: X = [feature(d) for d in dataset]  
        y = [float(d[5]) for d in dataset]
```

```
In [8]: y = tf.constant(y, shape=[len(y),1])
```

```
In [9]: K = len(X[0])
```

Note that we convert y to a native tensorflow vector. In particular we convert it to **column** vector. We have to be careful about getting our matrix dimensions correct or we may (accidentally) apply the wrong matrix operations.

Code: Gradient Descent in Tensorflow

Next we write down the objective – note that we use native tensorflow operations to do so

```
In [10]: def MSE(X, y, theta):  
         return tf.reduce_mean((tf.matmul(X, theta) - y)**2)
```

Next we setup the variables we want to optimize – note that we explicitly indicate that these are **variables** to be optimized (rather than constants)

```
In [11]: theta = tf.Variable(tf.constant([0.0]*K, shape=[K,1]))
```

```
In [12]: optimizer = tf.train.AdamOptimizer(0.01)
```

```
In [13]: objective = MSE(X,y,theta)
```

Initialized to zero

Stochastic gradient descent optimizer with learning rate of 0.01

Specify the objective we want to optimize – note that no computation is performed (yet) when we run this function

Code: Gradient Descent in Tensorflow

Boilerplate for initializing the optimizer...

```
In [14]: train = optimizer.minimize(objective) ← We want to minimize the objective
```

```
In [15]: init = tf.global_variables_initializer()
```

```
In [16]: sess = tf.Session()
sess.run(init)
```

Code: Gradient Descent in Tensorflow

Run 1,000 iterations of gradient descent:

```
In [17]: for iteration in range(1000):  
         cvalues = sess.run([train, objective])  
         print("objective = " + str(cvalues[1]))
```

```
objective = 7836.5107  
objective = 7836.5107  
objective = 7836.5107  
objective = 7836.5107  
objective = 7836.5103  
objective = 7836.5107  
objective = 7836.5107  
objective = 7836.5107  
objective = 7836.5107  
objective = 7836.5107  
objective = 7836.5107  
objective = 7836.5103  
objective = 7836.5103  
objective = 7836.5093  
objective = 7836.5093  
objective = 7836.5093  
objective = 7836.5093  
objective = 7836.5093  
objective = 7836.5093
```

Code: Gradient Descent in Tensorflow

Print out the results:

```
In [18]: with sess.as_default():  
         print(MSE(X, y, theta).eval())  
         print(theta.eval())
```

```
7836.5093  
[[ 0.23223479]  
 [-0.89481604]  
 [ 0.11925128]  
 [-0.4959688 ]]
```

Summary

Note that in contrast to our "manual" implementation of gradient descent, many of the most difficult issues were taken care of for us:

- No need to compute the gradients – tensorflow does this for us!
- Easy to experiment with different models
- Very fast to run 1,000 iterations, especially with GPU acceleration!

Other libraries

Tensorflow is just one example of a library that can be used for this type of optimization. Alternatives include:

- Theano - <http://deeplearning.net/software/theano/>
 - Keras - <https://keras.io/>
 - Torch - <http://torch.ch/>
 - Etc.

Each has fairly similar functionality, but some differences in interface

Questions?