Designs and Patterns for Parallel Programming

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Overview

Your presenter: Joe Hummel, PhD

- PhD in field of high-performance computing
- An exciting time to be working in this area...



- □ A few high-level design problems...
- □ *Pipeline and Dataflow*
- Using the Concurrent Data Structures provided by the TPL
- □ Producer-Consumer
- MapReduce and Task Local State
- □ Parallel LINQ
- □ Speculative Execution
- □ Asynchronous Programming Model

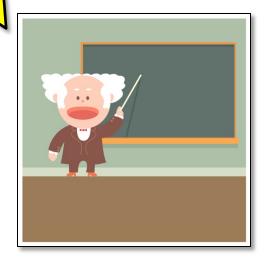


Design Problem #1

Your app needs to perform 100+ CPU-intensive operations, each taking roughly 3-5 minutes.

Execution order doesn't matter.

How do you execute the 100+ operations?



Create 100 tasks, 1 per op, with no creation options.

[Since long-running, .NET will over-compensate and create 100 worker threads; since CPU intensive, system will thrash.]

Create 100 tasks, 1 per op, with long-running option.

[.NET creates 100 dedicated threads; since CPU intensive, system will thrash.]

Create 1 task per core, as one finishes create another.

[use WaitAllOneByOne pattern, or MaxDegreeOfParallelism.]

Design Problem #2

Your app needs to perform 20+ I/O operations
— e.g. web page downloads. Order doesn't
matter, but downloads can take anywhere from
a few ms to a few seconds.

How do you download the web pages?



Create 1 code task per download, no special creation options.

[Starts one download for each worker thread in pool.]

Create 1 façade task per download using FromAsync + APM pattern (web object's Begin/End methods).

[Starts download and returns thread to pool, potentially allowing all 20+ downloads to start.]

Create 1 task per core, explicitly or via MaxDegreeOfParallelism.

[Starts one download per core.]

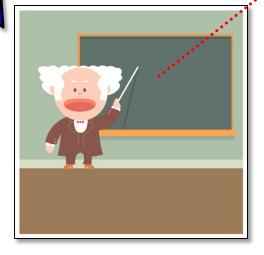
Use Producer-Consumer pattern, e.g. dedicating 1-2 tasks to download & remainder to processing.

[Consider when download time << pre>processing time.]

Design Problem #3

You want a logging task that runs for the duration of your app, logging say every 30 seconds.

What does the design look like?



This is harder than it looks:-)

Things to consider in your design:

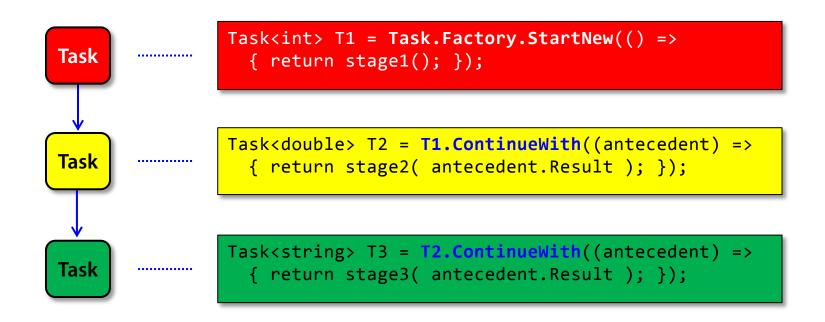
- 1. Create with long-running option
- 2. Before app closes you need to join with task and catch any exceptions
- 3. Design clean shutdown of task so app closes (use TPL cancellation?)
- 4. If task can crash, design a way to monitor and restart (check task's Status property via a timer, or app's message loop?)

Patterns

- Pipeline
- Dataflow
- Concurrent data structures
- Producer-Consumer
- Map-reduce
- Parallel LINQ
- Speculative Execution
- Asynchronous Programming Model

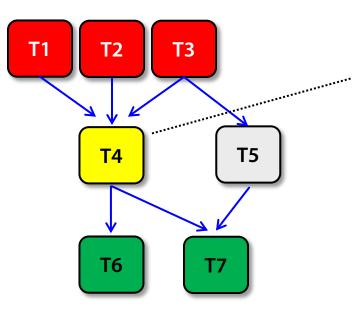
Pipeline

- Linear flow from one task to another
 - □ image processing, UI updating, workflows...



Dataflow

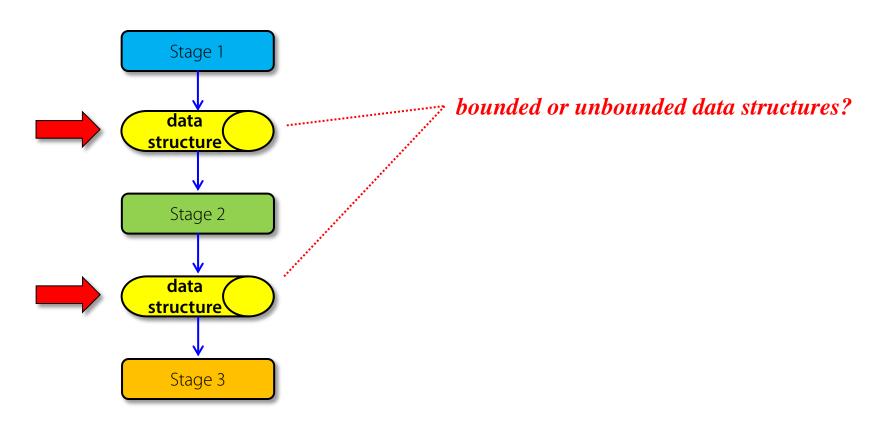
Generalized flow with many-to-one and one-to-many...



```
Task<int> T1 = ...;
Task<int> T2 = ...;
Task<int> T3 = ...;
Task<double> T4 =
  Task.Factory.ContinueWhenAll<int, double>(
     new[] {T1, T2, T3},
     (tasks) =>
        double result;
        foreach(Task<int> t in tasks)
        return result;
  );
```

Increasing parallelism

Thread-safe data structures allow stages to run independently...



Concurrent Data Structures

- TPL offers a set of thread-safe data structures:
 - □ ConcurrentBag<T>
 - □ ConcurrentQueue<T>
 - □ ConcurrentStack<T>
 - □ ConcurrentDictionary<T>





Example: ConcurrentQueue

Enqueue?

□ safe to call in parallel...

Dequeue?

- □ typically you check then dequeue...
- this is *still* unsafe: you cannot separate check from dequeue, some other task could grab in-between yielding race condition!
- □ Solution? API redesigned with TryDequeue method that checks and dequeues in one call

Enumeration?

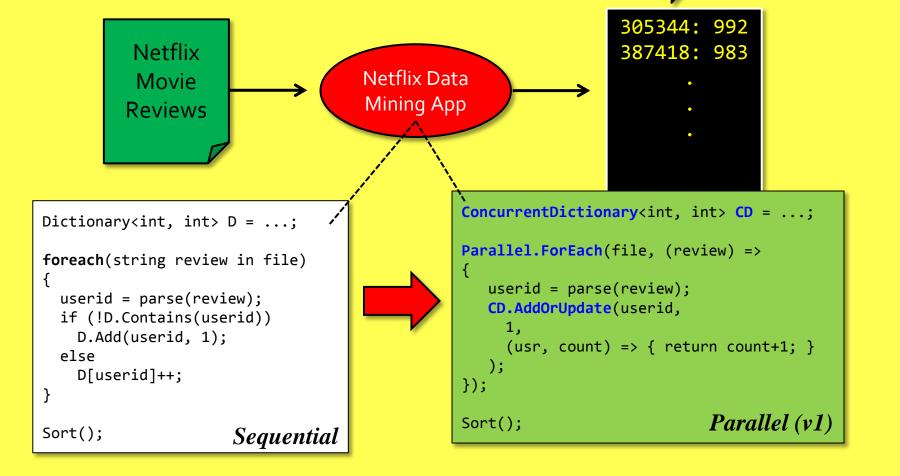
- □ Foreach, ToArray, ...
- □ thread-safe by creating snapshots

```
using System. Collections. Concurrent;
var Q = new ConcurrentQueue<int>();
Task.Factory.StartNew(() =>
  { Q.Enqueue(123); });
Task.Factory.StartNew(() =>
  { Q.Enqueue(456); });
Task.Factery.StartNew(() =>
    if (0.Cc at > 0)
      DoWor ( ). Dequeue() );
  });
Task.Factory.StartNew(() =>
  { Work w;
    if (Q.TryDequeue(out w))
      DoWork(w);
  });
Task.Factory.StartNew(() =>
  { contents = Q.ToArray(); });
```

DEMO

Users with most reviews (the "top 10" users)

Netflix data mining using a ConcurrentDictionary...



Results?

Netflix app worked correctly with ConcurrentDictionary



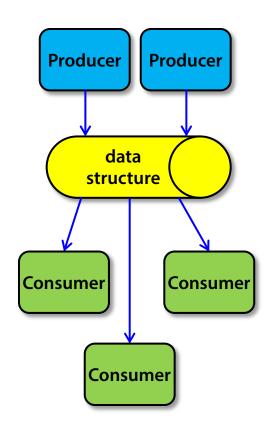
Unfortunately, parallel version ran slower



- \Box In this application, minimal data processing ==> contention for dictionary
- □ ConcurrentDictionary does internal synchronization / locking
- □ Too much contention ==> slower execution...

Producer-Consumer

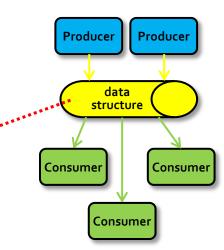
- Good pattern to use for long-running workloads where data generation speed is very different from data consumption speed
 - □ Data structure throttles faster component...
- Example:
 - producer(s) read from disk / network
 - consumer(s) process the data



BlockingCollection<>

- Think fixed-sized collection in a parallel world:
 - blocking producers if collection is full
 - blocking consumers if collection is empty

- Fixed-sized collections are more realistic
 - □ ... and typically *improve* performance by throttling faster component!



Implementation

```
Task producer = tf.StartNew(() =>
{
   for(...) //blocks task if Q is full:
     workQ.Add( work );

   //Signal we're done:
   workQ.CompleteAdding();
}
);
```

```
Task consumer = tf.StartNew(() =>
{
   while (!workQ.IsCompleted)
      try { // blocks task if Q is empty:
        T work = workQ.Take();
          // process work item:
      // Exceptions ==> no more work:
      catch(ObjectDisposedException)
      catch(InvalidOperationException) { }
```

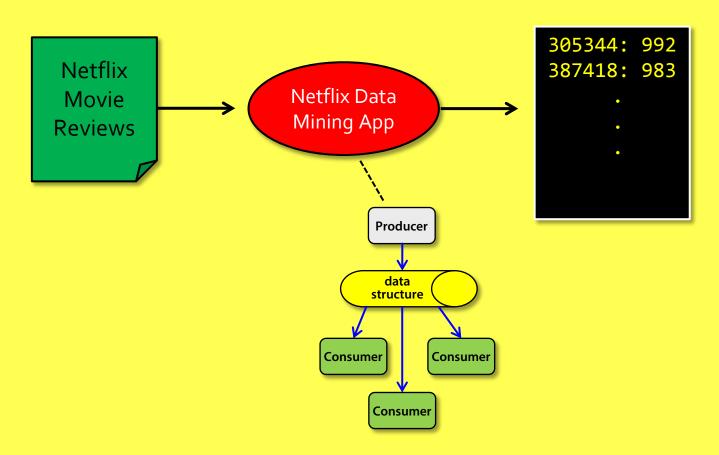
Producer |

data structure

Producer

DEMO (v2)

Netflix data mining using Producer-Consumer...

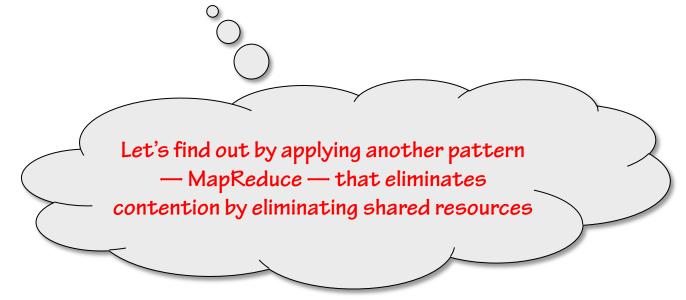


Results?

Producer-Consumer was correct *and* yielded a performance improvement!

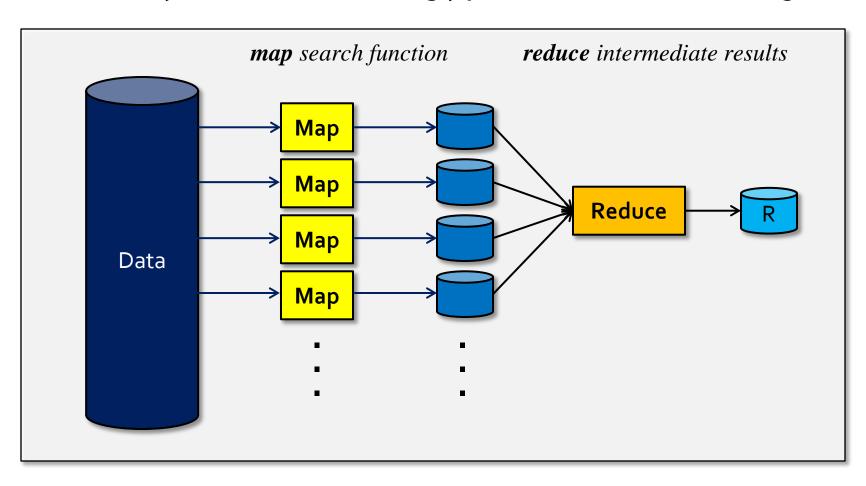


- However, improvement was sub-linear
 - □ Bad fit given that I/O time and processing times are roughly the same (fast)?
 - □ Contention for the blocking collection?



MapReduce

Commonly used for embarrassingly parallel search / data-mining...



Implementing MapReduce

- Various strategies...
 - □ fire off N Task<T>, use WaitAllOneByOne to reduce results as they finish
 - □ use Parallel.For / .Foreach with Task Local Storage (TLS)

```
for (int i=0; i<N; ++i)
  tasks.Add(Task.Factory.StartNew<T>(
      (data) => { return map(data); }
  );

while (tasks.Count > 0) // wait for tasks to finish:
{
  int index = Task.WaitAny( tasks.ToArray() );
  reduce(tasks[index].Result);
  tasks.RemoveAt(index);
}
```

```
Parallel.ForEach(datasource,
  () => { return new TLS(); }, // initializer:

  (datum, ..., tls) => // task body:
  {
    map(datum, tls);
    return tls;
  },

  (tls) => { reduce(tls); } // finalizer:
  );
```

DEMO (v3)

Netflix data mining using MapReduce...

```
305344: 992
                                                            387418: 983
            Netflix
                                   Netflix Data
            Movie
                                   Mining App
           Reviews
                                            Dictionary<int, int> result = new Dictionary<int, int>();
Dictionary<int, int> D = ...;
                                            Parallel.ForEach(File.ReadLines(infile),
foreach(string review in file)
                                              () => { return new Dictionary<int, int>(); },
                                              (line, loopControl, localD) =>
                                                                                    // task body:
 userid = parse(review);
 if (!D.Contains(userid))
   D.Add(userid, 1);
                                                userid = parse(line);
 else
                                                if (!localD.Contains(userid))
                                                                                           same as
   D[userid]++;
                                                  localD.Add(userid, 1);
                                                                                       :.... sequential
                                                else
                                                                                           version
Sort();
                   Sequential
                                                  localD[userid]++
                                                return localD;
                                              },
                                              (localD) => { lock(result) { merge(result, localD); } }
                                            );
```

Results?

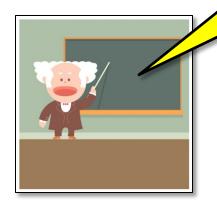
MapReduce yields best performance so far!



□ Eliminating contention usually does...

Extra credit for the viewer!

Take a look at the provided solution "Netflix-Par-FileChunk-LessStrict". It uses MapReduce but improves performance by partitioning the data itself, *and* also by a willingness to accept the occasional imprecise result.



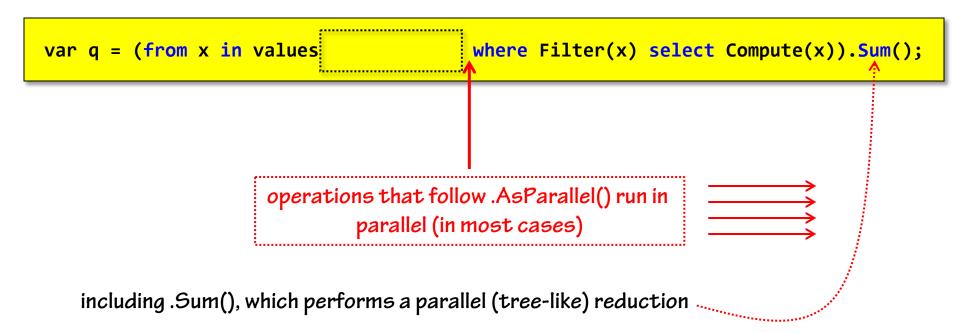
Parallel LINQ

- Are you a fan of LINQ Language Integrated Query?
- Then you'll love PLINQ and its support for:
 - □ Parallelism
 - □ *MapReduce*

```
var query = ( from x in values.AsParallel()
where Filter(x)
select Compute(x) ).Sum();
```

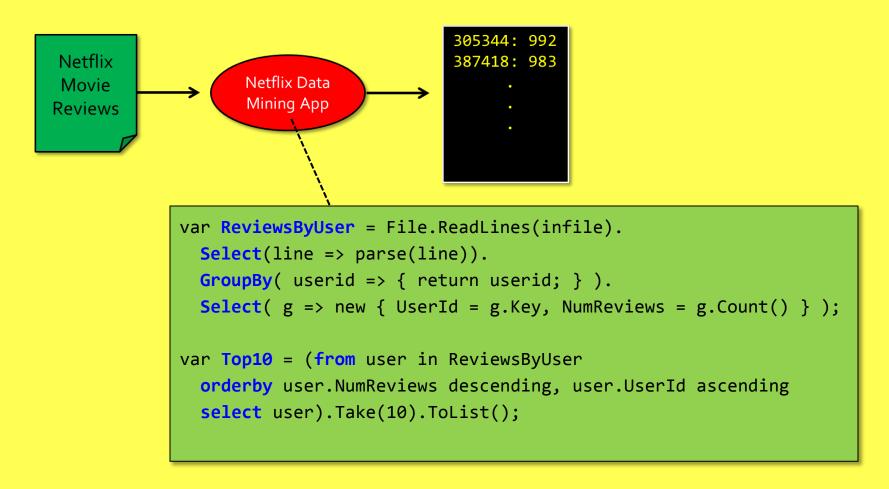
A brief overview...

- LINQ enables querying of IEnumerable data sources
- PLINQ turns an IEnumerable into a ParallelQuery data source
 - □ for in-memory collections



DEMO (v4)

Netflix data mining using PLINQ...



Speculative Execution

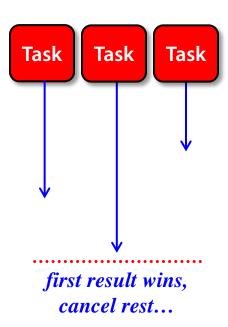
- When you have multiple sources for generating a result...
- Example:
 - □ Calling web services...

```
var cts = new CancellationTokenSource();
var token = cts.Token;

for (int i=0; i<N; ++i)
   tasks.Add(Task.Factory.StartNew<T>(
        () => { ... },
        token
   );

int winner = Task.WaitAny( tasks.ToArray() );
var result = tasks[winner].Result;

cts.Cancel(); // tell the rest to cancel...
```



Asynchronous Programming Model

- APM pattern is commonly used for asynchronous operations
 - □ File I/O, network I/O, ...
 - □ "**Begin**" starts operation / "**End**" completes operation & harvests result
- Various .NET classes support the APM pattern
 - □ FileStream: BeginRead / EndRead
 - □ HTTPWebRequest: BeginGetResponse / EndGetResponse
 - . . .

Advantage?

 NET classes start operation on a thread, but then return thread to pool until operation completes

TPL support for APM pattern

- TPL provides first-class APM support via façade tasks
 - □ Use FromAsync to create task that wraps calls to Begin / End
 - Use standard task mechanisms to wait / continue / harvest result

```
FileInfo fi = new FileInfo(filename);
int bytes = (int) fi.Length;
byte[] buf = new byte[bytes];
FileStream fs = new FileStream(filename, FileMode.Open, FileAccess.Read,
                               FileShare.Read, bytes, true /*async*/);
var T1 = Task<int>.Factory.FromAsync(fs.BeginRead, fs.EndRead, buf, 0, bytes, null);
var T2 = T1.ContinueWith((antecedent) =>
 {
    fs.Close();
     int bytesRead = T1.Result;
     . // process data in buf:
  });
```

DEMO

Historical stock data analysis...

```
>> Please enter stock symbol(s) (e.g. 'msft,intc,...'): aapl,intl,msft

** CSV Stock History App [any-cpu, release] **
    Stock symbol(s): 'aapl,intl,msft'
    Time period: last 10 years
    Internet access? True

** aapl **
    Data source: 'http://nasdaq.com, daily Close, 10 years'
    Data points: 2,515
    Min price: $6.56
    Max price: $100.69
    Std dev/err: 97.964 / 1.953

** intl **
    Data source: 'http://moneycentral.msn.com, weekly Close, 1 year'
    Data points: 53
    Min price: $13.87
    Max price: $22.21
    Std dev/err: 3.452 / 0.474

** msft **
    Data source: 'http://moneycentral.msn.com, weekly Close, 1 year'
    Data points: 53
    Min price: $23.27
    Max price: $23.27
    Max price: $23.27
    Max price: $25.73
    Std dev/err: 1.362 / 0.187

** Done **

Press any key to continue . . .
```

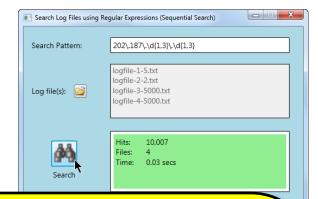
Good example of both patterns:

- 1. Speculative execution of downloads
- 2. APM and façade tasks for HTTP requests

Aside: Parallel I/O

- I/O is *hard* to parallelize
 - requires support from underlying
 OS and physical hardware
 - □ sensitive to data sizes

don't expect much performance from commodity hardware

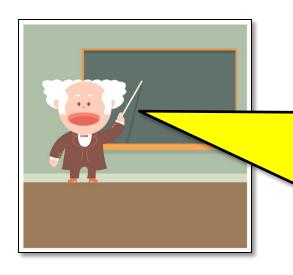


Design Challenge!

The first lecture — "Understanding the Dangers of Concurrency" — focused on a Log File Search app.

However, none of the provided solutions work for arbitrarily large files (they all fail with an "Out of Memory" exception).

Provided in the "before\" sub-folder of this lecture is a sequential version that works for arbitrarily large files, along with two parallel versions that fail (one does synchronous I/O, and the other does async I/O using the APM). Based on what you've learned in this course, design a correct, high-performing parallel version for arbitrarily large files.



Summary

- Software design is challenging
- Asynchronous and Parallel software is even more challenging :-)
- Success measured in terms of both correctness and performance
- Increase your odds of success by using patterns:
 - □ Structured parallelism
 - □ *Pipleline and Dataflow*
 - □ Concurrent data structures
 - □ Producer-Consumer
 - □ *MapReduce*
 - \Box *PLINQ*
 - □ Speculative execution
 - \Box APM

References

- Microsoft's main site for all things parallel:
 - http://msdn.microsoft.com/concurrency
- LINQ:
 - LINQ: http://code.msdn.microsoft.com/101-LINQ-Samples-3fb9811b
 - □ PLINQ: http://msdn.microsoft.com/en-us/library/dd46o688.aspx
 - Parallel.Foreach vs. PLINQ:

http://blogs.msdn.com/b/pfxteam/archive/2010/04/21/9997559.aspx

- I highly recommend the following short, easy-to-read book:
 - Parallel Programming with Microsoft .NET: Design Patterns for Decomposition and Coordination on Multicore Architectures, by C. Campbell, R. Johnson, A. Miller and S. Toub, Microsoft Press

Online: http://tinyurl.com/tpl-book