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Do Work in Progress (WIP) - Limit in Agile Software Development Matter?

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Abstract

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Preface

Part I

Introduction

Chapter 1

Background

In this master thesis the main topics will contain analyzing of the gathered data from Software Innovation (SI) in order to see if WIP - limit in agile methods matters. SI is a Scandinavian software company that delivers Enterprise Content Management applications. From 2008 to 2013 SI gathered information about each task that was developed. The main reason SI gathered the data was to see if Kanban was more sufficient than Scrum for their use. For interested reader, the case study can be found in the article "Quantifying the Effect of Using Kanban versus Scrum: A Case Study" (Sjøberg, Johnsen and Solberg, 2012). In this thesis the data will be used to determine if WIP-limit matters in agile methods.

In this chapter there will be a brief introduction of Scrum, Kanban, affiliated tools and to Software Innovation.

1.1 Lean

The Lean approach was introduced between 1948 and 1975 in manufacturing work in Japan. It was designed to find and eliminate waste, so the manufacturing could deliver value to the customer more efficiently. In 2003 Mary and Tom Poppendieck first introduced Lean thinking to software development, they published 'Principles of Lean Thinking' (M. Poppendieck and T. Poppendieck, 2003). Poppendieck stated that an important tool to manage work flow is the concept of pull-systems, which means tasks are put in production only when a customer asks for it (M. Poppendieck and T. Poppendieck, 2009). In the recent years, Kanban has been introduced more and more to software development, and is becoming one of the keys to Lean practice in the field (Anderson et al., 2011).

1.2 Kanban

"We can define Kanban software process as a WIP limited pull system visualized by the Kanban board" (Anderson et al., 2011).

Toyota production system introduced Kanban as a scheduling system for lean and just-in-time (JIT) production during late 1940's and early 1950's in order to catch up with the American car industry. The Kanban method combined with the lean approach was a success for Toyota. The success was noticed by the software development industry among other. In the last ten years more and more, software development companies have started to implement agile methods such as Scrum and Kanban (Conboy, 2009), (Ohno, 2001).

More and more software projects adapt to Kanban, and this is one of the reasons why this thesis will focus on Kanban and one of its main tools WIP. Kanban is one of the agile method in the wind these days, and is used with Lean Software development which is one of the fastest growing approaches in software development (Anderson et al., 2011) One of the most important people in Kanban software development, David Anderson also referred to as "father of Kanban in the software development industry" (Gupta, 2013) and author of the book "Kanban: Successful Evolutionary Change for Your Technology Business" once stated "If you think that there was Capability Maturity Model Integration, there was Rational Unified Process, there was Extreme Programming and there was Scrum, Kanban is the next thing in that succession." (Leonardo Campos, 2013)

The Kanban method splits one big problem into many small pieces of problems. When the small pieces are defined by the team, the problems are put up on the Kanban-board

to visualize the problems and see potential bottlenecks during development. When people start to understand Kanban, they easily discovered where the bottlenecks are (Shinkle, 2009).

Kanban system focus on:

- Continuous flow of work
- No fixed iterations or sprints
- Work is delivered when it's done
- Teams only work on few tasks at the time specified by the WIP limit
- Make constant flow of released tasks (Anderson et al., 2011).

One of the main difference between Scrum and Kanban is estimation, in simulation of software maintenance process, with and without a work-in-process limit (Concas et al., 2013) estimation was defined to be the main source of waste. In their research, they find out, if they let the developers work with small tasks at time and not be interrupted, they will be more effective. The developers in this case was interrupted when they was assigned to estimate tasks. The research groups decided to implement Lean-Kanban, which includes minimizing waste, which meant estimation for this case. After implementing Lean-Kanban the teams increased the ability to perform work, lower the lead time and meet the production dates.

1.2.1 Kanban Board

"The Kanban board makes it clear to all the team members the exact status of progress, blockages, bottlenecks and they also signal possible future issues to prepare for"(Middleton and Joyce, 2012).

The Kanban board is one of many important tools in Kanban. It's used to control the WIP, increase the information flow with visualization and spot bottlenecks (Concas et al., 2013). A Kanban is board is illustrated in figure 1.1. Each column has an intuitive name in order to describe itself so the developers easily can track where each task is. In figure 1.1 each column is named "Backlog", "In progress" and "Done". Each column can have a WIP-limit to specify how many works in progress there are allowed in the column (Middleton and Joyce, 2012). In figure 1.1 the WIP-limit is stated under the column name. The backlog columns have a WIP-limit of 4, In progress has 5 and obviously done doesn't have a WIP-limit. The yellow stickers represents the tasks. Some follow to path to mark stickers with different colors representing the severities. In "Kanban Implementation in a Telecom Product Maintenance" the stickers has three different colors, green, yellow and red. Each of the color indicates how close to overdue the tasks are. If the sticker is red, the tasks is already overdue (Seikola, Loisa and Jagos, 2011).

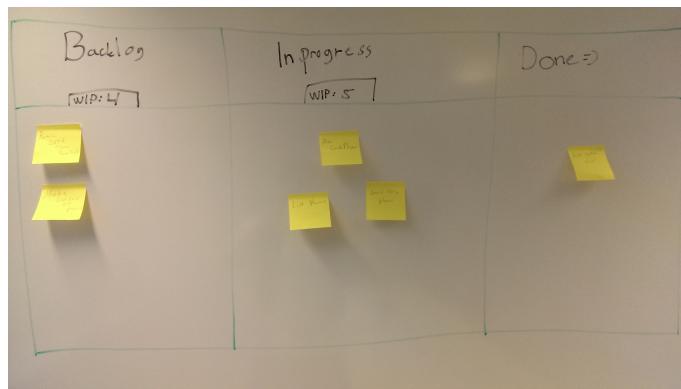


Figure 1.1: Example of a Kanban board

1.2.2 WIP-limit

"WIP-limits seem to be the worst understood part of the Kanban system. When used properly, it exposes bottlenecks and reduces lead time for individual work items. Used improperly, it can starve developers for work or result in too many people working on the same work items." (Shinkle, 2009)

WIP-limit is a tool in Kanban to reduce overhead by limit task-switching for each developer and visualize bottlenecks. One of the best way to explain WIP and the impact of WIP-limit is to use cars and roads as analogy. All roads have its maximum capacity of cars. When this limit is reached, traffic jam occurs and the throughput of cars decreases and lead time increases. The same can be said about software development teams, a software team has a maximum number of tasks they can perform, if the team is pushed over the maximum limit, the throughput of tasks decreases and lead time increases.

Research in the field of Work In Progress (WIP) and if WIP-limit matters in software development lacks proper research. But some research has been done, Giulio Concas and Hongyu Zhang for instance has done research on the difference between limit WIP and unlimited WIP(Concas et al., 2013) and David Anderson, Giulio Concas, Maria Ilaria Lunesu, and Michele Marchesi has also highlighted the difference between limit WIP and unlimited WIP the article 'Studying Lean-Kanban Approach Using Software Process Simulation' (Anderson et al., 2011). Lukasz proposes to use the effectiveness metric to help determine WIP-limit. The effectiveness metric should be applied after end cycle according to Lukasz, so after each cycle, one can apply it and the result could be used as a guideline for WIP-limit for the next cycle. The effectiveness takes the number of bugs found (a_i) and the number of bugs found by external people (e.g. lawyers, accountants, coaches, consultants, translators, internal and external service providers ,etc.) (e_i), and minus a_i and e_i , then divide the result by a_i and multiply it by 100% as shown in 1.1 (Sienkiewicz, 2012)

$$Ei = \frac{ai - ei}{ai} * 100\% \quad (1.1)$$

In manufacture business Taho Yanga, Hsin-Pin Fub, Kuang-Yi Yanga stated that WIP could be defined as:

$$WIP - limit = cycle time * throughput rate \quad (1.2)$$

(Taho Yanga and Yanga, 2007).

When first implementing Kanban, Shinkle explains that the users don't care about WIP or setting a WIP-limit, but rather the visibility of Kanban through the Kanban board. When the user get more experience with Kanban, they start to attempt the principles of WIP-limit (Shinkle, 2009). Srinivasan, Ebbing and Swearing said that setting the WIP-limit is not easy. They suggest that the WIP-limit is set, and then observe throughput, and adjust after that (Srinivasan, Ebbing and Swearingen, 2003).

The principle of Kanban also says you should limit WIP, but what should the limit be? The principles of Kanban tells us to experiment (Kniberg, 2010). Lean Software Management and the Impact of Kanban on Software Project Work suggest that WIP should be minimized as well. The study suggest to minimize WIP-limit to keep high quality (Ikonen et al., 2011) and to create continuous flow and bring problems to the surface (Middleton and Joyce, 2012). The conclusion of present study is to keep the WIP-limit low and experiment by slowly increase the WIP-limit until the throughput decreased and lead time increased, then you know that the previous WIP-limit was the perfect one.

1.2.3 Benefits with WIP-limit according to various articles

1. When lowering the WIP-limit will help people avoid task switching. When one is task switching it's hard to be able to fully concentrate. (Ikonen et al., 2011).
2. There's stated when using short-cycle times and Kanban board to limit WIP, the software development team's learning is increased' (Middleton and Joyce, 2012)
3. The team in Lean Software Management study realized the where bottlenecks. So the team started to determine WIP by their constraints. The team quickly realized that they had fewer quality assurance/testing staff and business analysts than software developers. This reflected the bottlenecks and the constraints, so the team adjusted the WIP-limit to how much work they could handle; this gave the team more experience in dealing with WIP and increased productivity (Middleton and Joyce, 2012).

4. To set a WIP-limit has some advantages:

- It helps team to reduce overhead
- Decrease lead time
- Increase throughput
- It reduces flow times
- Reduces variation
- Improves quality

(Mark L. Spearman and J.Hopp, 1990).

As one can see, several people have their opinion on WIP and the importance of defining WIP-limit. If we disregard Lukasz's effectiveness metric, there is no clear rule on how to determine the WIP-limit without recording throughput and lead time and measure them, even though WIP is a crucial tool in order to use Kanban sufficient.

1.2.3.1 Limit WIP vs. Unlimited WIP

Simulation of software maintenance process, with and without a work- in-process limit did a research on how the throughput and how developers experience WIP limit and unlimited WIP (Concas et al., 2013).

One of the result from this paper was at the end of a simulation, the average of closed tasks was 4145 when the WIP was limited and 3853 when the limit was not limited (about 7% less). The paper concludes their finds; developers are more focused on fixing few issues, because the number of issues they can work on is limited. The developers are more likely to continue on the issue from the day before, rather than starting on another issue, this reduce overhead. When developers start on a new issue, they need to use time to familiarize themselves with the code and the issue. That could create unnecessary overhead if some developer already has done it, but that developer is now working on a another issue. The study also showed that limit WIP can improve throughput and work efficiency. (Concas et al., 2013).

1.3 Waterfall

1.4 Scrum

The Scrum framework is the source of much of the thinking behind principles and values of the Agile Manifesto. Values as "Individuals and interactions over processes

and tools", "Working software over comprehensive documentation", "Customer collaboration over contract negotiation" and "Responding to change over following a plan" relates directly to Scrum. (Alliance, 2012).

Scrum have three main roles, the Product Owner, the Scrum Master and the members of the development team. The Product owner in collaboration with the Scrum Master decides which work to be prioritized in the backlog. A backlog represents the tasks to be done in order to complete the project. The Scrum Master acts like a team leader and helps the team and organization to take best advantages of Scrum. The development team works on tasks specific for the sprint there in (Alliance, 2012).

Sprint is a time-boxed interval over a given time. The Scrum framework suggests the duration of sprints to be from one to four weeks. Before each sprint, a sprint planning meeting is conducted, with all the team members attending. A Sprint planning meeting is held so the team can discuss tasks from the backlog and come to an agreement of which tasks to be put in the minimal backlog. (Alliance, 2012).

In each sprint a minimal backlog is created so the developer knows which tasks to work on. The Product Owner and the team members discuss and decide which tasks from the backlog to be added to the minimal backlog. After the minimal backlog is full, the Product Owner and the team members discuss each task in order to get a better and a shared understanding of what is required in order to complete the task. One of the main principles in Scrum is that it requires that a new feature is ready for release after a sprint. The feature should be a visible part of the product in order to get feedback from end-users. So all the tasks in the minimal backlog combined should be a visible of the product. (Alliance, 2012).

1.5 Lead time

"Lead time is the total elapsed time from when a customer requests software to when the finished software is released to the customer" (Middleton and Joyce, 2012).

Lead time is an essential ingredient when you look for the optimal WIP. Often in project, lead time is split into pieces, so every task has its own lead-time; this gives the development teams the advantages to experiment with different WIP's in order to see the different lead-times and then measure which WIP that suits this project the best.

The citation by Joyce above is close to definition of what lead time is. This definition could be useful for consultancy companies, but for in-house development company with few releases each year this definition is unsuitable. Quantifying the effect of Using Kanban versus Scrum stated two reason why this is unsuitable for in house

development companies:

"First, the amount of time a work item remains in the backlog queue before it's put on the board is a function of priority, not whether the company uses Scrum, Kanban, or other development methods. Furthermore, companies that develop and sell products to many customers might propose new features themselves and put them on the backlog before any customers request them. Second, given a policy of two or three releases a year, the result of a work item isn't delivered to the customer immediately after it's finished" (Sjøberg, Johnsen and Solberg, 2012).

1.6 Just-In-Time

"Just-In-Time is based on delivering only the necessary products, to the necessary time and the necessary quantity." (Lai, Lee and Ip, 2003).

Just-In-Time was introduced 30 years ago by Toyota Motor in combination with Lean. JIT has been developed to increase productivity through waste reduction and increasing the value added on the production processes. In one of the books by Mary and Tom Poppendieck the JIT principle is explained (M. Poppendieck and T. Poppendieck, 2006). To explain, illustrate and visualize the JIT principle Mary and Tom Poppendieck uses the picture 1.2 (Lai, Lee and Ip, 2003) (M. Poppendieck and T. Poppendieck, 2006).

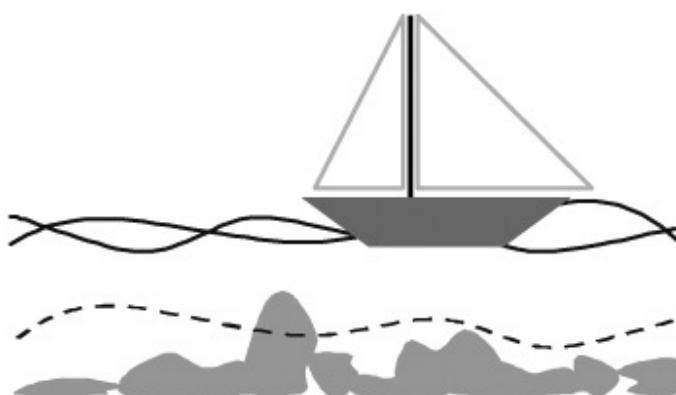


Figure 1.2: JIT example

In the picture 1.2 the stream reflects the inventory. Under the stream, there are rocks located in different sizes. The rocks illustrates waste and problems that can occur. If the stream level is lowered, the rocks are more visualized. At this point you have to clear out rocks in order to make the boat continue it's journey, or it will crash into the rocks. After the rocks are cleaned out, you can lower the stream level again and continue until

it's just pebbles left

If one lower the stream, problem and waste will become visible. But why do Lean want to lower inventory in order to make problems and waste occur? Because when problem and waste occurs, you are able to fix the problem and remove the waste. Fixing the problem and removing the waste can have several benefits such as, your process could be optimized and you are one step closer to have zero problems and zero waste. (Lai, Lee and Ip, 2003) (M. Poppendieck and T. Poppendieck, 2006).

1.7 Throughput

"The output of a production process (machine, workstation, line plant) per unit time (e.g., parts per hour) is defined as the systems throughput or sometimes throughput rate" (Adams and Smoak, 1990)

The main concept of throughput is to measure how productive teams, people or companies are. Throughput is measured in number of finished delivered tasks per hour, day, week, month, quarter or year. This applies to both software development and manufacturing. But there are some differences. In software development each task is more abstract than in manufacturing and in software development each task can have different solutions depending on how the team or developer approaches the task.

In manufacturing each task usually has one solution and when the solution is found the physical item is mass produced. Adam said "Throughput in plant, line or workstation, is defined as the average quantity of good parts produced per unit time" (Adams and Smoak, 1990), which gives a good example of the relationship between tasks in manufacturing and software development. In manufacturing the part either fits its purpose (good) or not (defective). In software development a task can fit a purpose, but the purpose may be wrong.

A key factor in successfully measuring throughput in software development is to specify the size of each task. If the size is not specified a developer x can have throughput of 1, but another developer could have throughput of 3 and they still have done the same amount of work and this will give wrong results if the throughput is measured. So it's recommended to specify the amount of work for each task in advance.

1.7.0.2 Example of throughput measurement

This is a simple example to illustrate throughput with different task sizes. Team x had a throughput of eighteen tasks after the first quarter, twenty after the second, fifteen after the third and twelve after the last quarter and they used Scrum the first two quarters and Kanban the last two as illustrated by table 1.1. It will look like team x benefits most from Scrum. But if the task during the Kanban time was twice the size of Scrum, Kanban would suite team x the best. In order to get valid result from throughput measurement the size of tasks has to be agreed upon by the teams or company.

Quarter	Throughput	Method
1	18	Scrum
2	20	Scrum
3	15	Kanban
4	12	Kanban

Table 1.1: Throughput

1.8 Churn

"Churn is defined as the sum of the number of lines added, deleted, and modified in the source code" (Sjøberg, Johnsen and Solberg, 2012).

Churn is a measure that's not so quite known as lead time, throughput or WIP. Churn is a term that's used as surrogates for effort in software engineering. Many studies in software engineering use code churn or revisions as surrogate measure of effort (Yamashita, Anda, Mockus et al., 2012). Emam stated that "analysts should be discouraged from using surrogate measures, such as code churn, unless there is evidence that they are indeed good surrogates (El-Emam, 2000)." The authors in "Quantifying the effect of code smells on maintenance effort" agree that one should cautious when using surrogates for effort.

1.9 Software Innovation (SI)

Software Innovation is a Scandinavian software company. SI develops and delivers market-leading Enterprise Content Management applications that helps organizations

improve and increase efficiency in document management, case handling and technical document control. SI build products around Microsoft Sharepoint platform and tightly integrated into the Microsoft Office environment (Sjøberg, Johnsen and Solberg, 2012). (*Software Innovation* 2013).

SI has approximately 300 employees. SI has offices in Oslo, Copenhagen and Stockholm and a development center in Bangalore(*Software Innovation* 2013). From 2001 to 2006 SI used Waterfall process with a life cycle of design, implementation, testing, and deployment for each new release. In 2007 SI changed to Scrum and in 2010 SI went from Scrum to Kanban.

Part II

The project

Chapter 2

Research Questions

In this thesis the overall research question will be to study the effects of WIP limits for SI. Hopefully the research will give answers to the following question:

- See if there exist an optimal WIP-limit for a given context.
- How to best find the optimal WIP-limit
- Which parameters to consider in order to optimize WIP.

Chapter 3

Research Methods

In this chapter the research methods will be revealed and the reason why the data from Software Innovation is picked for this master thesis. The first section 3.1 will be about the choice of case and complementary information about SI. The next section 3.2 reveals the algorithm of how the program measure WIP for each day. The next section 3.3 will provide a comprehensive example of how the program measure WIP per day. The consecutively sections reveals the algorithm of how the program measure bugs (section 3.4) throughput (section 3.5), churn (section 3.6) and lead time (section 3.7) per day. In the last section information about SPSS will be given (section 3.8)

3.1 Choice of case

The case study in this thesis is based on data set from an in-house software company named Software Innovation (SI). The data set contains information about each task SI has worked on in the last five years (2008-2013). The data set is represented in a excel document, a excerpt of the document is shown in table 3.2.

In order to try answering the research questions stated in chapter 2 the data from SI will be interpreted. The data set contains thirty columns with different data for each task, mostly of this columns are irrelevant for this study but the important columns is stated in table 3.4. The reason SI and the data set from SI is analyzed in this thesis is based on the article "Quantifying the Effect of Using Kanban versus Scrum" by Dag Sjøberg, Anders Johnsen and Jørgen Solberg (Sjøberg, Johnsen and Solberg, 2012) since Dag is the supervisor of this thesis and had access to the data set.

From 2001 to 2006 SI used the Waterfall process with a life cycle of design,

implementation, testing, and deployment for each new release. In 2007 SI changed to Scrum. Scrum was implemented with the standard elements of Scrum:

- Cross functional teams
- Sprint planning meetings
- Estimation of work items using planning poker
- Daily standup meetings
- Sprints of three weeks

After a three week sprint a fully tested shippable code was ready. In 2010 SI went from Scrum to Kanban. SI felt that Scrum was too rigid and didn't fit their purpose, they also feared that inaccurate estimation and time boxes gave them longer lead times. They also saw Scrum planning meetings as waste which reduced productivity and quality(Sjøberg, Johnsen and Solberg, 2012).

SI decided to implemented Kanban in the following manner. When a work item starts SI tries to make the item flow through all the stages until it's ready for release. This procedure happens as quick as possible. In order for an item to be ready for release it has to be at a satisfactory quality level, which is defined by SI. SI also implemented WIP limits, if the WIP limit is reached, no new tasks are started until another task is finished also known as just-in-time. (Sjøberg, Johnsen and Solberg, 2012).

In this thesis the main goal is to see if WIP matters in agile method. It will be done by interpret the data set from SI on team level. SI has about 25 teams. From the data set there are six teams who have worked on the most tasks. These six teams will be used in this thesis in order to try answer the research question in section 2.

The data from SI will be analyzed using a program made for this thesis, which will extract the variables shown in table 3.1 for each of the six team. This data will be further analyzed using SPSS. With SPSS each variable will be gathered based on which quarter of the year it belongs to and team. The analyzed data will be presented in chapter 4.

The following tables (table 3.1 and table 3.2) shows the five variables that will measured by the program and the latter shows excerpt from the data set

- WIP
- Throughput
- Bugs
- Lead time
- Churn

Table 3.1: Variable

ID	Type	Created Date	From Day	Date To	Lead Time	Team
3027	Bug	2008-10-07	2008-10-09	2008-10-16	20	Team one
3028	Bug	2008-10-07	2008-10-07	2008-10-08	10	Team six
3029	Feature	2008-10-07	2008-12-30	2008-12-30	105	Team two
3030	Feature	2008-10-07	2008-10-07	2008-10-07	1	Team three
3035	Bug	2008-10-08	2008-11-20	2008-11-28	17	Team five
3037	Feature	2008-10-08	2008-10-19	2008-10-19	7	Team three
3040	Bug	2008-10-10	2008-11-19	2008-11-19	48	Team one

Table 3.2: Excerpt from the data set

In the two following subsection the important columns from the data set will be shown and a brief description will be given (subsection 3.1.1) and the next subsection contains a table of which columns needed to calculate the variable stated in table 3.1 In the following table (table 3.3) information about day, date and quarter is stated.

- The Date standard is specified as yyyy-mm-dd.
- All seven days in the week are taken into account when measuring included Saturdays and Sundays
- Quarter of a year is defined as:
 - January, February and March (Q1)
 - April, May and June (Q2)
 - July, August and September.(Q3)
 - October. November and December (Q4)

(Investopedia, 2013)

Table 3.3: The standard of the data set

3.1.1 The columns

Column	Description
Created Date	When a task is put in backlog
Date From	When a given task is taken from the backlog
Date to	When a task is done. Done is defined by SI to be ready for release.
Lead Time	The amount of time elapsed from the date the task was created until the tasks has finished
Process Type	States the process used by the team which contains Kanban or Scrum
Type	The type column is labeled as either bug or feature depending on the type of the task
Lead time	Number of days used on a tasks
Lines added	Number of lines added to a feature or bug
Lines modified	Number of lines modified when working on a feature or bug
Lines deleted	Number of lines deleted from a bug or feature
Team	States the team who has been working on the task.

Table 3.4: Information about the columns from the SI dataset

3.1.2 Columns needed for calculating the variables per day

Variable	Description	Columns from SI
WIP	The number of items in progress on the given day	Date From and Date To.
Throughput	Number of tasks finished on a given day	Date To
Churn	Lines added, lines modified and lines deleted added together	Lines added, lines modified and lines deleted
Bugs	The number of tasks labeled as Bug and not feature	Type
Lead time	The time used on a task	Lead time

Table 3.5: Information about the columns from the SI dataset

3.2 WIP per day

3.2.1 Step 1: Gather all unique dates into a ArrayList

First step of WIP measurement is adding every date in the date from column into a ArrayList. Before the date is added, the program checks if the date is already in the ArrayList as shown in line two in listing 3.1. If the date contains in the ArrayList, then the date has already been measured, so the program will discard the date. The ArrayList will contain WIP objects, which will contain information about each date. The information about the object is stated in table 3.6 ¹

Type	Variable name
Date	start
Date	end
String	team
String	processType
int	WIP

Table 3.6: Variables of the WIP objects

Pseudocode step 1:

```

1 for date IN date_from_column:
2     if date not contains in Arralist

```

¹ArrayList is a resizable array implementation of a list. The ArrayList class provides function for manipulate the size of the array, check the size of the list and convert the list to an array (Oracle, 2013)

```

3     WIP = nr_of_date_occurrence(date)
4     ArrayList.put(WIP)
5
6 int nr_of_date_occurrence(Date date)
7 for d IN date_from_column DO
8   if d EQUALS date DO
9     nr_of_date_occurrence++
10  return nr_of_date_occurrence
11

```

Listing 3.1: Gather all unique dates into ArrayList

3.2.2 Step 2: Gather the remaining dates

The data set from SI contains dates from 2008 to 2013, but there are some dates missing which table 3.2 show. In table 3.2 the date 2008-10-09 is missing. In order to generate WIP for each day the program has to create the dates that's not in the set.

In order to create the remaining dates, the program takes the first date and the last date from the ArrayList created in previous section (3.2.1) as shown in listing 3.2 line one and two. Then the program checks if all the dates between the first date and the last date are in the ArrayList. If the dates are not in the ArrayList, the program will put the date into the ArrayList as show in method on line ten to thirteen. In order to keep the pseudocode simple, the generateWIP method stated in line twelve was omit. The method creates a new WIP object and returns it.

```

1 WIP first = ArrayList.get(0)//points to the first WIP object in the ArrayList
2 WIP last = ArrayList.get(ArrayList.size())//points to the last WIP object in
   the ArrayList
3 Next_date //points to the next date
4 Next_date = first.getDate() // Next_date assigned before iteration
5 while Next_date NOT EQUALS last.getDate()
6   New_date = Next_date + 1 //Finds the next date
7   AddToArraylist(New_date)
8   Next_date = New_date
9
10 addToArraylist(Date d)
11   if d NOT CONTAINS IN ArrayList
12     WIP = generateWIP(d)
13     ArrayList.add(WIP)
14

```

Listing 3.2: Gather the remaining dates

3.2.3 Step 3 Measure WIP

The ArrayList from section 3.2.1 and 3.2.2 now contain a WIP object for each date from 2008 to 2013. The reader needs remember the WIP counter of every WIP objects contains the number of tasks started on the corresponding date. This measurement was done in section 3.2.1. In this step the program will loop through the ArrayList, during the iteration each WIP object is extracted from the ArrayList and the new WIP is measured based on how many tasks that has been finished illustrated in the listing 3.3.

```
1 lastWIP = 0
2 CurrentWIP = 0
3 for WIP Object IN ArrayList
4   CurrentWIP = WIP.getWIP()
5   Nr_of_finishedDates = Occurrence_of_date(WIP.getDate())
6   WIP_measured = CurrentWIP - Nr_of_finishedDates + lastWIP
7   WIP.setWIP(WIP_measured)
8   WIP.setDate(WIP.getDate())
9   newArrayList.add(newWIP)
10  currentWIP = WIP_measured
11
12 Occurrence_of_date(Date date)
13   for d in The Date To column
14     if date AFTER d DO
15       Nr_of_dates_to_decrement++
16   return Nr_of_dates_to_decrement
17 %% Her mangler jeg å skrive at jeg setter isPicked slik at datoer ikke blir
measured flere ganger, burde jeg ta med det?
18
```

Listing 3.3: Meausre WIP

3.3 Example

This section will provide a comprehensive example of how the program WIP algorithm works. Figure 3.1 shows tasks id in the y-axis and dates in the x-axis. The green line indicates the duration of the task. The figure visualize how many work in progress (WIP) there are. Such as on the date 2008-10-12, tasks 3, 5 and 6 are in progress, which means the WIP is 3 for 2008-10-12.

In the example dates from figure 3.1 and from table 3.7 will be used to illustrate how the algorithm measure WIP

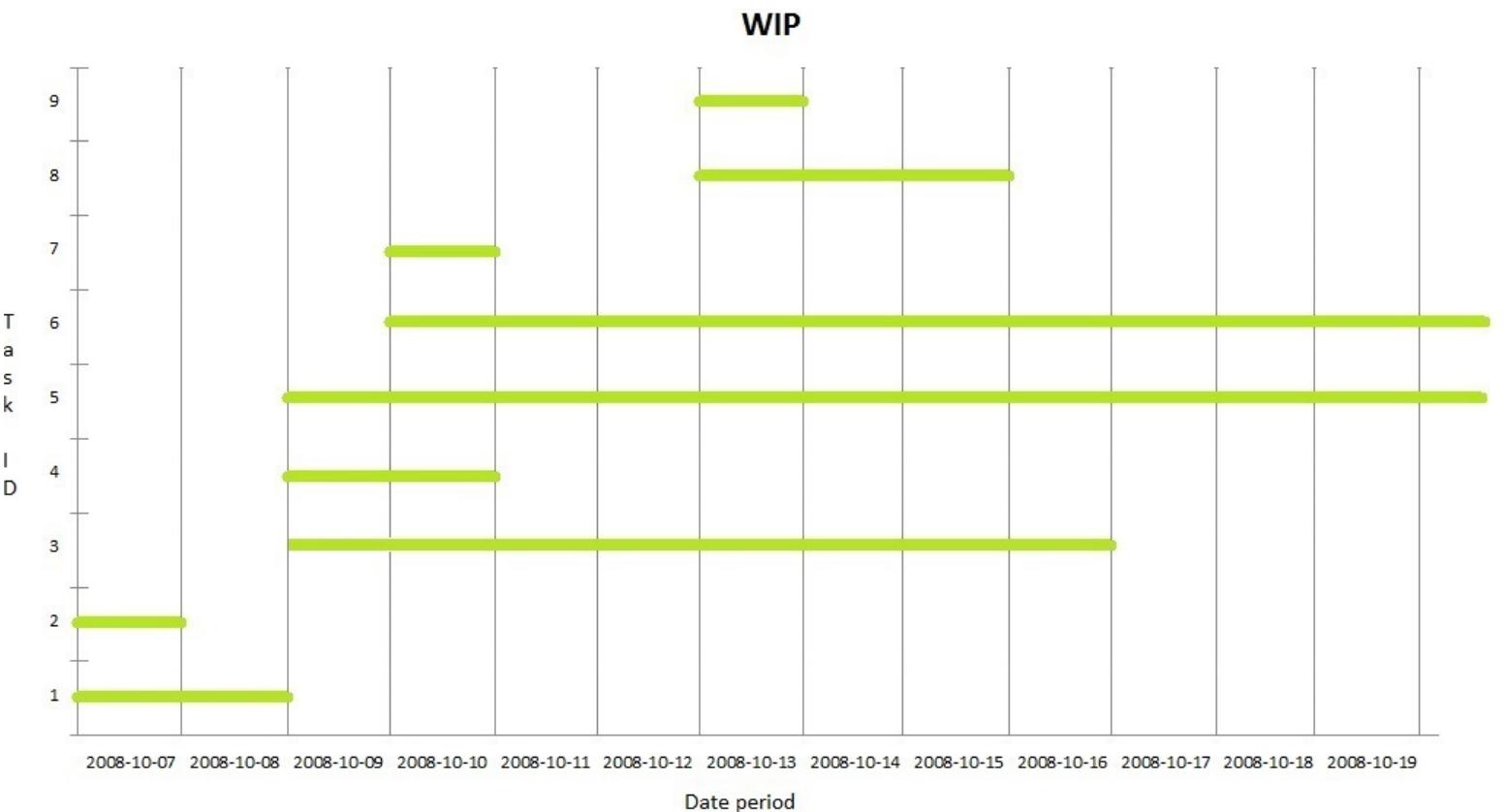


Figure 3.1: Illustrating the WIP timeline for this example

Task ID	Date From	Date To
1	2008-10-07	2008-10-08
2	2008-10-07	2008-10-07
3	2008-10-09	2008-10-16
4	2008-10-09	2008-10-10
5	2008-10-09	2008-11-04
6	2008-10-10	2008-11-05
7	2008-10-10	2008-10-10
8	2008-10-13	2008-10-15
9	2008-10-13	2008-10-13

Table 3.7: Showing Task ID, Date From and Date to

3.3.1 First step

First the algorithm will do a look-up on the date 2008-10-07 referred with task ID one and two in table 3.7 and measure the current WIP counter, which is two, because its two tasks in progress on this date. After this the program will put the date and the current WIP into a ArrayList. Now the algorithm will see if any task was done in the last period, since task one and two is the two first tasks to be measured in this example, there's no task finished in the period and there's no current WIP from other tasks, so the current WIP at 2008-10-07 is two illustrated by figure 3.1.

3.3.2 Second step

The program will do a look-up on the date 2008-10-09 referred with task ID three, four and five in table 3.7 and measure the current WIP for this date, which is three. Then, the program will put the date and the current WIP into a ArrayList. Next, the program will see that two tasks were done in the period of 2008-10-07 to 2008-10-09. So the program will measure that three new tasks were started at 2008-10-09, two where finished and the WIP from last date measure where two, so the calculation of WIP at 2008-10-09 is $3 - 2 + 2$, which gives a WIP of three on 2008-10-09 illustrated by figure 3.1.

3.3.3 Third step

The program will do a look-up on the date 2008-10-10 referred with task ID six and seven in table 3.7 and measure the current WIP for this date, which is two. Next the program will put the date and the current WIP into a ArrayList. Last the program will see that no task where done in the period 2008-10-09 to 2008-10-10 and the currently WIP is three, so this gives a new WIP of 5 illustrated by figure 3.1.

3.3.4 Fourth step

The program will do a look-up on the date 2008-10-13 referred with task ID eight and nine in table 3.7 and measure the current WIP for this date, which is two. Then the program will put the date and the current WIP into a ArrayList. Next the program will look at the period between 2008-10-10 and 2008-10-13, and see that task four and seven was done in this period. The WIP at 2008-10-13 will be $2 - 2 + 5 = 5$ as illustrated by figure 3.1. As illustrated by the example, WIPs are not decrement until the finished date

is passed, even though the task is done on the date, it's also worked on the same date, therefore the WIP is decrement after each date is passed.

3.4 Bug

Each task in the data set is labeled as either feature or bug as shown in figure 3.8. When the program reads in the data file, each task labeled as bug are saved in a data structure. The code is shown in listing 3.4.

Task ID	Type
57970	Bug
57971	Bug
57972	Bug
57973	Bug
57974	Feature
57975	Feature
57976	Feature
57977	Feature
57978	Feature

Table 3.8: Example of how tasks are labeled

```
1 void findBug()
2     while inputFile != EOF // End Of File
3         newLine = readLine()
4         if newLine.Type EQUALS 'Bug'
5             B = New Bug()
6             B.startDate = newLine.startDate
7             B.process = newLine.process
8             B.team = newLine.team
9             AddNewBug(B)
```

Listing 3.4: Pseudocode example of how bugs are found

3.4.1 Add Bug

When adding a new bug, each bug is gathered into a data structure based on which team the bug belongs to, as the tests in lines 2, 7, 12 and 17 of the listing 3.5 shows. After the right team is found the program tries to add the bug to the corresponding data structure as illustrated in lines 3, 8, 13 and 18. If the date of newly arrived bug

already contains in the data structure, a counter representing the date is incremented and the new bug is discarded as shown the method `dateExists` in lines 26 to 32.

Since the program knows which team the new bug belongs to (after the checks on line 3, 8, 13 and 18) a counter can represent each bug. In our analyze the only important for us is to know how many bugs there are on a given date and which team it belongs to.

```

1 void addBug(Bug b)
2   if b.team EQUALS "360"
3     if dateExists(b.date, 360.dataStructure) EQUALS false
4       // if date don't exists, then add the bug
5       360.dataStructure.add(b)
6
7   if b.team EQUALS "Neon"
8     if dateExists(b.date, Neon.dataStructure) EQUALS false
9       // if date don't exists, then add the bug
10      Neon.dataStructure.add(b)
11
12  if b.team EQUALS "Frontend"
13    if dateExists(b.date, Frontend.dataStructure) EQUALS false
14      // if date don't exists, then add the bug
15      Frontend.dataStructure.add(b)
16
17  if b.team EQUALS "Krypton"
18    if dateExists(b.date, Krypton.dataStructure) EQUALS false
19      // if date don't exists, then add the bug
20      Krypton.dataStructure.add(b)
21
22 void dateExists(Date d, DataStructure structure)
23   for Bug b in structure
24     if b.date EQUALS d
25       b.counter++
26       return true
27
28 return false
29

```

Listing 3.5: Pseudocode example of how bugs are added

3.5 Throughput

Finding the throughput per day is quite similar to how bugs are found (described in section 3.4). When reading in the data set a new throughput object is created for each line in the data set. Then all throughput objects are sorted based on team association. When all throughput objects are sorted, the program measure throughput. The throughput measurement is similar to the dateExists method (lines 26 to 32) in the listing 3.5 stated in section 3.4.1.

The dateExists method starts off with a test, the same test is done for bugs. If the date of the throughput object is in the data structure the corresponding counter is incremented. If the date is not in the data structure, the new throughput object is added to the data structure. An excerpt of the code is listed in 3.6

```

1 dateExists(Throughput tp d, DataStructure structure)
2   for Throughput t in structure
3     if t.date EQUALS tp.date
4       t.counter++
5       return
6
7 structure.add(tp);
8

```

Listing 3.6: Pseudocode example of how throughput is measured

3.6 Churn

As stated in section 1.8 in order to take churn into account one need to know it's good surrogates. SI has gathered churn with help of Microsoft's Team Foundation Server (TFS). The TFS system automatically records data such as churn and lead time. Based on TFS one can know that churn for SI a is good surrogate.

To measure churn the data set from SI contains three columns as shown in table 3.9. To complete the measure the three columns are multiplied. For task id one, the churn is 2028 ($352 + 307 + 1369 = 2028$). The task with id six has zero churn. In the data set there exists tasks with zero churn, these tasks don't need code in order to be completed such tasks needs support to be finished.

Task id	Lines added	Lines modified	Lines deleted
1	352	307	1369
2	314	31	15
3	314	31	15
4	62	327	153
5	21	3	0
6	0	0	0

Table 3.9: How churn is presented in the excel document

3.7 Lead time

The program does not need to analyze the lead time for each task. The lead time for each task is stated in the set as shown in table 3.10. The program will gather all the tasks that are started on the same day and belong to the same team and add up their lead time together as shown in code listing 3.7.

ID	Type	Leadtime
84096	Feature	1
84118	Bug	25
84096	Feature	7
84118	Bug	13

Table 3.10: How lead time is recorded in the excel document

```
1
2 addLeadTime(lead_time t, DataStructure structure)
3   for lead_time in structure
4     if lead_time.date EQUALS t.date
5       lead_time.leadtime+= t.leadtime
6     return
7   structure.add(t)
```

Listing 3.7: Pseudocode example of lead time is measured

3.8 SPSS

3.8.1 Introduction

After the program has analyzed the data from SI, SPSS will be used on the data derived from the program. SPSS will help to answer the research question stated in chapter 2 with help of these three statistics method: Correlation, Graph's and case summaries. The case summaries will be used on:

- WIP
- Throughput
- Lead time
- Churn
- Bugs

3.8.2 SPSS

"IBM® SPSS® Statistics is a comprehensive system for analyzing data. SPSS Statistics can take data from almost any type of file and use them to generate tabulated reports, charts and plots of distributions and trends, descriptive statistics, and complex statistical analyses." (IBM, 2014)

From SPSS the function stated in figure 3.2 will be used to analyze data.

- Case summaries
- Line Graph
- Correlation

Figure 3.2: SPSS functions used to analyze data in this thesis

3.8.2.1 Case summaries

The case summarize function is used on WIP, throughput, lead time, bugs and churn to calculate in variable in figure 3.3 per quarter:

- Mean
- Median
- Standard Deviation
- Number (N)
- Max
- Min

Figure 3.3: Descriptive statistic

3.8.2.2 Line Graph

The line graph function is used to line graph with both one and two y-axis.

3.8.2.3 Correlation

The correlation function is used to look at the relationship between variables. In this thesis the correlation function will help determine if there are any relationship between WIP, throughput, bugs, lead time and churn. In order to determine the relationship between variables the variables correlation goes from -1 to 1.

- If the correlation is between ± 0.4 it's a low correlation
- If the correlation is between $\pm 0.5 - 0.8$ it's a medium correlation
- If the correlation is between ± 0.8 it's a strong correlation
- If the correlation is positive it means that the variables are decreasing or increasing at the same time.
- If the correlation is negative it means that the variables are going in the opposite direction.
- If the correlation is low, it means that there is no detective pattern between the variable and there is no relationship exists between the variable.
- If the correlation is high, it means that the exists a pattern between the variables.

Chapter 4

Results

4.1 Introduction

In this chapter

4.2 Team 1

4.2.1 WIP correlation

As shown in table 4.1 bugs, throughput, churn, churn bugs and throughput bugs have a significant correlation with WIP. In table 4.2 is descriptive statistic for WIP shown. As shown by the N column all quarters besides 2010-3 give reasonable data since in each quarter has 90 days give or take.

Correlation	Bugs	Throughput	Churn	Churn Bugs	Throughput bugs
WIP	.753*	.637**	.738*	.708*	.800**

Table 4.1: Team one - Correlation - WIP

* Correlation is significant at the 0.05 level (2-tailed).

** Correlation is significant at the 0.01 level (2-tailed).

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
360	2010-3	39	5.82	5	2.55	15	1
	2010-4	92	1.7	2	0.70	4	1
	2011-1	90	4.37	2	6.87	31	1
	2011-2	91	14.25	5	14.497	52	3
	2011-3	92	2.4	3	0.97	4	1
	2011-4	92	14.26	4	22.718	97	1
	2012-1	91	13.35	5	16.62	67	4
	2012-2	91	15.35	3	27.145	94	2
	2012-3	92	16.97	13.5	8.719	52	5
	2012-4	90	9.07	2	17.073	74	1
Total		860	9.99	4	16.04	97	1

Table 4.2: Descriptive statistics for WIP - Team one

4.2.1.1 WIP and throughput correlation

Both throughput bugs have a significant correlation relationship with WIP. This is inconsistent with Kanban philosophy and what the research says about WIP-limit. All the research and the principles of Kanban suggest lowering WIP in order to get high throughput. In team one's case the higher the WIP the higher the throughput.

In table 4.2 the mean column shows indication that team one could have experimented with WIP-limit the first five quarters. But in the last quarter, the WIP-limit decreased. In the mean column in table 4.3 if one take a look at the quarter 2012-4, the throughput are second highest, the same quarter as WIP decreased. This link could have several reasons. One could be that they experimented with WIP and found out after the quarter 2012-3 that they needed to lower the WIP in order to increase the throughput, which it did. Or maybe team one didn't care about setting a WIP-limit as stated in section 1.2.2 by Shinkle.

The throughput bugs variable and throughput variable both have a significant correlation with WIP, which make sense since throughput bugs are a subset of throughput. But throughput feature on the other hand has a -0.441 correlation with WIP. As shown in table 4.6. The correlation number -0.441 is not a significant correlation, but it's on the opposite site level of throughput and throughput feature although it's also a subset of throughput. This also goes in the literatures favor, because when WIP is low, more throughput feature is produced. Maybe this could have something to do with the relationship bugs and feature? Features are often more challenging and more fun to work with than bugs. One could also argue that maybe it's was significant fewer tasks labeled as bug than feature or vice versa, but as shown in table 4.6 that's not the case.

Team	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
360	2010-3	5	2.6	2	1.34	4	1
	2011-1	7	5.71	6	4.03	13	1
	2011-2	26	3.38	3	1.96	8	1
	2011-3	1	1	1	.	1	1
	2011-4	15	6.13	4	4.984	18	2
	2012-1	16	6.56	6.5	3.59	14	1
	2012-2	13	10.15	10	5.742	23	1
	2012-3	22	3.41	2	3.33	12	1
	2012-4	14	8.14	7	6.5	21	1
	Total	119	5.55	4	4.694	23	1

Table 4.3: Descriptive statistic for throughput - team one

Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
2010-3	14.00	3.07	2.00	3.47	14.00	1.00
2010-4	71.00	2.73	2.00	2.05	11.00	1.00
2011-1	67.00	2.55	2.00	1.96	13.00	1.00
2011-2	38.00	1.97	2.00	1.03	4.00	1.00
2011-3	43.00	1.65	1.00	0.92	4.00	1.00
2011-4	53.00	1.83	1.00	1.31	6.00	1.00
2012-1	40.00	1.90	1.00	1.46	7.00	1.00
2012-2	27.00	2.07	2.00	1.36	6.00	1.00
2012-3	40.00	2.70	2.00	2.43	13.00	1.00
2012-4	3.00	1.00	1.00	0	1.00	1.00
Total	396.00	2.26	2.00	1.82	14.00	1.00

Table 4.4: Descriptive statistic for throughput feature - team one

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
360	2010-3	9	1.11	1	0.33	2	1
	2010-4	3	1	1	0	1	1
	2011-2	26	5	3	4.45	17	1
	2011-3	1	1	1	.	1	1
	2011-4	24	6.79	4	6.35	20	1
	2012-1	18	4.83	4	3.20	11	1
	2012-2	19	7.84	9	5.843	17	1
	2012-3	6	2.67	1.5	2.25	6	1
	2012-4	11	3.18	3	1.77	7	1
	Total	117	5.08	3	4.88	20	1

Table 4.5: Descriptive statistic for throughput bugs - team one

Number of features	Number of bugs
660	594

Table 4.6: Number of throughput divided into bugs and feature - Team one

4.2.1.2 WIP and bugs correlation

Only looking at WIP and throughput correlation without taking other factors into account may give bias.

Bugs and churn have also have a significant correlation with WIP. A positive bug correlation could help favor what said about WIP and throughput in literature. Since both WIP and throughput increases and decreases at the same time, the same goes for bug. This could mean that when WIP is high, and throughput is high team one also has a lot of bugs in their code. This means that they don't do their job well, and this could be because they have a lot of task switching due to high WIP-limit. As stated in table 4.7 one can see that almost each bug that was reported was finished in the same quarter. The outlier could be the bugs reported just before the quarter end. Worth noticing in quarter 2011-1 there was no bugs recorded. This conclude the statement that the WIP-limit is set too high. 0.467

Quarter	Percent of bugs finished in the same quarter as reported
2010-3	100%
2010-4	100%
2011-1	0%
2011-2	97.74%
2011-3	20%
2011-4	97.5%
2012-1	96.51%
2012-2	98.63%
2012-3	82.35%
2012-4	100%

Table 4.7: The percent number of bugs finished in the same quarter as reported - Team one

4.2.1.3 WIP and churn correlation

Churn and WIP have a positive correlation. This means that when team one has more tasks in progress the tasks requires more coding. This is odd and could be a coincidence. Churn and churn bugs have also here as with throughput a close correlation with WIP. But churn feature has a correlation of 0.467

As shown in table 4.8 and the column 'maximum' the biggest feature task requires almost 60 times more code editing than the biggest bug task as showed in table 4.9. In the same table the column std. deviation shows way more proliferation for churn feature. This indicates that the numbers in churn feature are way more spread, which again indicates that it's random of how big the tasks are. In the churn bugs the std. deviation is much lower than in the feature table. This could explain why churn feature not has a positive correlation as churn and churn bugs. Both median and mean favors that churn feature requires more work than churn bugs

Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
2010-3	5	434	106	808.67	1879	30
2011-1	7	0	0	0	0	0
2011-2	26	1199.54	139	4695.424	24139	0
2011-3	1	0	0	.	0	0
2011-4	15	516.4	148	1063.92	3946	0
2012-1	16	5775.25	410	20692.09	83317	0
2012-2	14	7569.14	522.5	24655.70	93108	0
2012-3	23	6.74	0	25.44	118	0
2012-4	16	408	47	1056.23	4275	0
Total	123	2001.29	65	11380.06	93108	0

Table 4.8: Descriptive Statistic for churn feature - Team one

Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
2010-3	7	57.14	22	73.91	193	4
2010-4	2	30	30	41.012	59	1
2011-2	21	386.33	177	431.94	1604	2
2011-3	1	2	2	.	2	2
2011-4	19	306.89	148	437.13	1557	12
2012-1	16	312.81	160	307.73	1035	4
2012-2	13	296.92	169	292.29	1019	18
2012-3	7	37.71	9	56.494	126	0
2012-4	8	118.75	84	103.09	292	3
Total	94	260.48	135	346.25	1604	0

Table 4.9: Descriptive Statistic for churn bugs - Team one

4.2.2 Throughput

In table 4.10 is correlation for throughput for team one stated. Throughput has a positive correlation to bugs, WIP, churn, churn bugs, churn feature, throughput bugs and churn average. In table 4.11 is descriptive statistic for throughput shown. The range from 2011-2 to 2012-4 is the most valuable quarters based on the number of tasks finished if one excludes the quarter 2011-3.

Correlation	Bugs	WIP	Churn	Churn Bugs	Churn feature	Throughput bugs	Churn Average
Throughput	.945**	.816**	.962**	.776**	.736*	.940**	.638*

Table 4.10: Team one - Correlation - Throughput

* Correlation is significant at the 0.05 level (2-tailed).

** Correlation is significant at the 0.01 level (2-tailed).

Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
2010-3	5	2.6	2	1.34	4	1
2011-1	7	5.71	6	4.03	13	1
2011-2	26	3.38	3	1.96	8	1
2011-3	1	1	1	.	1	1
2011-4	15	6.13	4	4.984	18	2
2012-1	16	6.56	6.5	3.59	14	1
2012-2	13	10.15	10	5.742	23	1
2012-3	22	3.41	2	3.33	12	1
2012-4	14	8.14	7	6.5	21	1
Total	119	5.55	4	4.694	23	1

Table 4.11: Descriptive statistic for throughput - team one

4.2.2.1 Throughput and bugs correlation

The correlation between throughput and bugs is 0.945, which is significantly high. The quarters from 2011-2 to 2012-4 minus 2011-3 is also the most valid in bugs. The correlation between bugs and throughput means that when team one has high throughput of tasks they also produce more bugs. This correlation can be used to favor the literature of WIP as stated in section 4.2.1. Since team one produces more bugs when they finished feature or bugs could be an indicator that they may have a too high WIP-limit so they cannot focus enough on each task.

Throughput and bugs also have the same relationship as WIP and bugs, because throughput and throughput bugs has a significant correlation to both, but throughput

feature not. Throughput feature have correlation of -0.340 compared to -0.441 in WIP. What this means is that when team one has high throughput of tasks, they mostly produces tasks labeled as bug. This could be because the feature tasks are more comprehensive than the tasks labeled as bug so the lead time is longer for feature tasks than bug tasks. Or it could due the reason that when there is an upcoming release, all the bugs are pushed through the Kanban system.

In table 4.12 the hypotheses that feature tasks is more comprehensive is verified. Four out of the six valid quarters churn feature is significant higher. In table 4.13 one can see that throughput bug's lead time is higher than throughput feature in four out of six quarters. The lead time cannot be counted, since bugs usually have longer lead time than feature, because feature usually has higher priority than bugs.

Quarter	Churn feature	Churn bug
2010-3	0	29.3
2010-4	0	20
2011-1	0	0
2011-2	1199.54	59.14
2011-3	0	2
2011-4	516.4	76.66
2012-1	5775.25	1009.9
2012-2	7569.14	707.46
2012-3	6.74	15.68
2012-4	408	168.28

Table 4.12: team one - Average churn for feature and bugs - Bugs

Quarter	Throughput feature	Throughput bug
2010-3	23	51.33
2011-1	38.85	6
2011-2	19.73	25.46
2011-3	1	5
2011-4	30.6	36.38
2012-1	20.37	45.56
2012-2	35.30	55.89
2012-3	30.72	17.67
2012-4	30.57	8.82
Total	27.30	35.09

Table 4.13: Lead time for throughput feature and bugs per quarter

4.2.2.2 Throughput and churn correlation

Churn, churn bugs, churn feature and the average of churn all have a positive correlation with throughput which mean that when throughput is high, the tasks requires more coding. From table 4.14 in the number column one can see that in each quarter there is a lot of tasks. Although from the median column one can see that in 6 of the valid quarters, three of them have a median of 0, which gives an indicator that in these three quarter there is probably minority of tasks which has high churn. So the correlation between throughput and churn may not be valid.

Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
2010-3	11.00	17.00	11.00	18.94	60.00	0
2011-1	73.00	0	0	0	0	0
2011-2	136.00	14.40	3.00	21.51	83.00	0
2011-3	48.00	1.67	0	7.71	50.00	0
2011-4	156.00	14.81	6.50	19.91	86.00	0
2012-1	157.00	9.47	0	19.55	86.00	0
2012-2	238.00	10.11	0	18.50	83.00	0
2012-3	53.00	3.02	0	9.62	53.00	0
2012-4	26.00	21.42	14.50	20.55	67.00	1.00
Total	902.00	10.14	0	18.55	86.00	0

Table 4.14: Descriptive statistic for churn - team one

4.2.3 Bugs

Correlation	Throughput	WIP	Churn	Churn Bugs	Churn feature	TP bugs
Bugs	.945**	.753*	.992**	.813**	.730*	.984**

Table 4.15: Team two - Correlation - Throughput

* Correlation is significant at the 0.05 level (2-tailed).

** Correlation is significant at the 0.01 level (2-tailed).

It's hard to find any correlation between lead time, WIP and throughput. In quarter three in 2011, lead time, WIP and throughput dropped to their lowest point. This was due to the summer vacation. In quarter two of 2011 and two of 2012 bugs dropped but both Throughput and WIP increased.

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
360	2010-3	7	1.29	1	0.48	2	1
	2010-4	4	1	1	0	1	1
	2011-2	32	4.16	3.5	3.63	14	1
	2011-3	5	1	1	0	1	1
	2011-4	26	6.15	4	5.80	22	1
	2012-1	21	4.09	4	2.7	11	1
	2012-2	17	8.59	9	5.59	19	1
	2012-3	7	2.43	2	1.512	5	1
	2012-4	11	3	3	2	6	1
	Total	131	4.53	3	4.44 t	22	1

Table 4.16: 360 - Descriptive statistic - Bugs

4.3 Team 2

4.3.1 WIP

For team two there is no significant correlation between WIP or any of the variables that have been measured. In table 4.18 one can see that every quarter beside 2010 is a valid quarter based on the number of dates in quarter. An excerpt of the correlation data is shown in table 4.17. As in section 4.2.1 for team one, throughput is fairly high, not significant high as for team one, but fairly high. For team two the correlation between bugs and WIP is quite low, so the conclusion stated in section 4.2.1 may not suit team two.

In table 4.19 the correlation for throughput for team two is shown. Here one can see that throughput and bugs have a significant correlation. That means when WIP increases, throughput increases and when throughput increases the number of bugs reported increases. Which could mean that the WIP-limit is to high.

	Throughput	Bugs	Churn	Leadtime
WIP	0.341	0.278	-0.446	0.096

Table 4.17: Team two - Correlation - WIP

* Correlation is significant at the 0.05 level (2-tailed).

** Correlation is significant at the 0.01 level (2-tailed).

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Frontend	2010-3	24	9.28	10	6.43	23	1
	2010-4	92	13.92	13	3.931	25	5
	2011-1	90	15.3	15.5	3.94	23	7
	2011-2	91	23.51	24	4.19	37	13
	2011-3	92	20.13	20	5.43	32	9
	2011-4	92	21.33	21	6.92	34	7
	2012-1	91	22.91	22	6.63	40	11
	2012-2	91	21.93	21	3.36	32	17
	2012-3	92	25.93	27	4.92	36	19
	2012-4	67	19.18	15	9.66	41	9
Total		822	20.18	20	6.93	41	1

Table 4.18: Descriptive statistic - WIP - Team two

Correlation	Bugs	Throughput feature	Bugs finished in the same quarter
Throughput	.885**	.669*	.640*

Table 4.19: Team one - Correlation - Throughput

* Correlation is significant at the 0.05 level (2-tailed).

** Correlation is significant at the 0.01 level (2-tailed).

4.3.2 Lead time

Correlation	Churn Bugs	Churn union	Precent bugs fininshed	Churn Average
Leadtime	-.771**	.704*	-.697*	.766**

Table 4.20: Team one - Correlation - Leadtime_union

Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
2010-3	14.00	3.07	2.00	3.47	14.00	1.00
2010-4	71.00	2.73	2.00	2.05	11.00	1.00
2011-1	67.00	2.55	2.00	1.96	13.00	1.00
2011-2	38.00	1.97	2.00	1.03	4.00	1.00
2011-3	43.00	1.65	1.00	0.92	4.00	1.00
2011-4	53.00	1.83	1.00	1.31	6.00	1.00
2012-1	40.00	1.90	1.00	1.46	7.00	1.00
2012-2	27.00	2.07	2.00	1.36	6.00	1.00
2012-3	40.00	2.70	2.00	2.43	13.00	1.00
2012-4	3.00	1.00	1.00	0	1.00	1.00
Total	396.00	2.26	2.00	1.82	14.00	1.00

Table 4.21: Descriptive statistic - Throughput - Team two

4.3.3 Analyzed data per quarter

In figure 4.1 the average WIP, Throughput, Lead time and Bugs per quarter for team 360 are shown. As stated in section 1.2.2, to get best possible throughput the WIP-limit should be low. According to the figure 4.1, when WIP is high, throughput is also high and when WIP is low, throughput also low. This team data is contrary to what is suggested from Kanban and research.

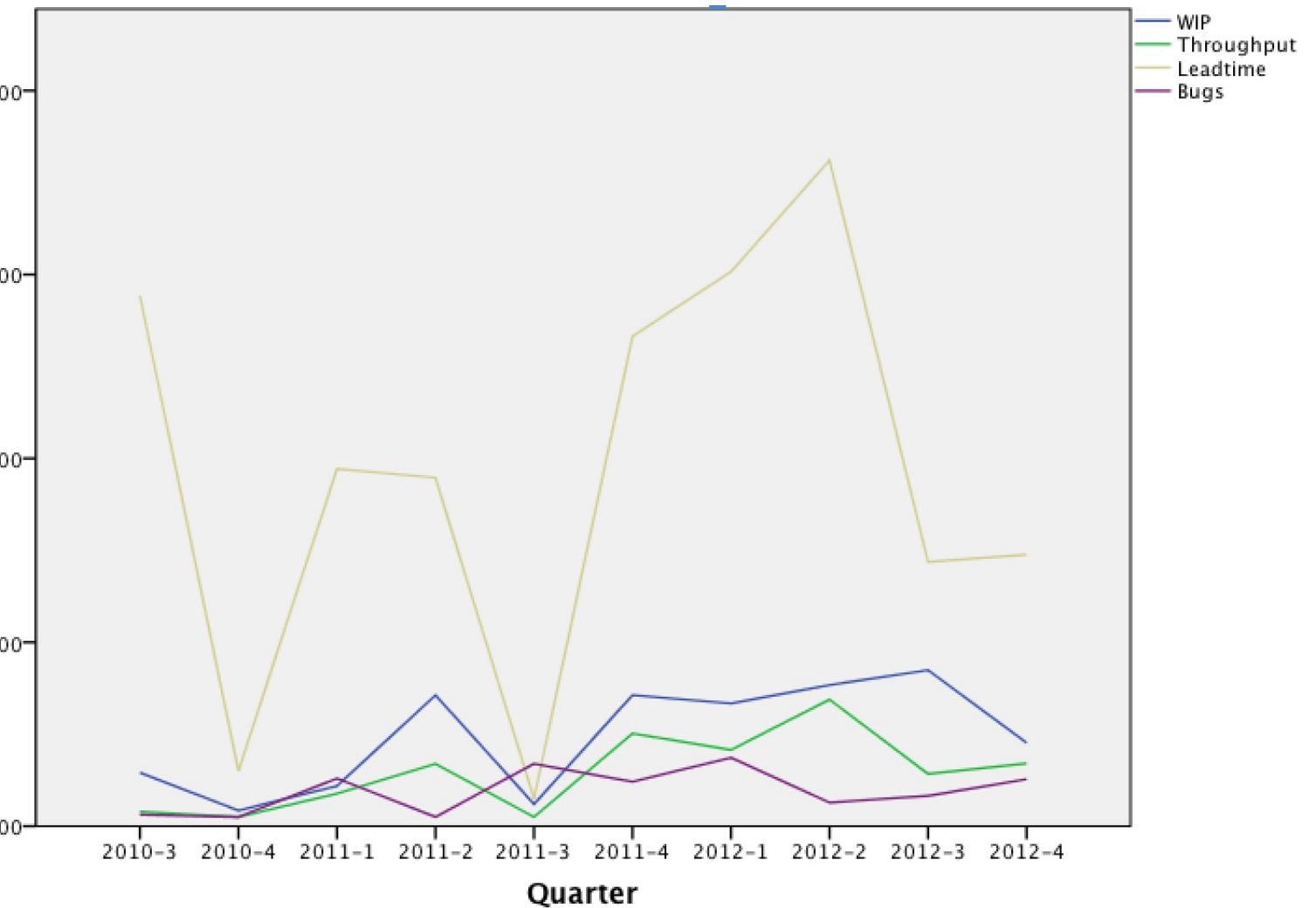


Figure 4.1: WIP, Throughput, Lead time and Bugs for 360 per quarter

In figure 4.2 the average WIP, Throughput, Lead time and Bugs per quarter for team Frontend are shown

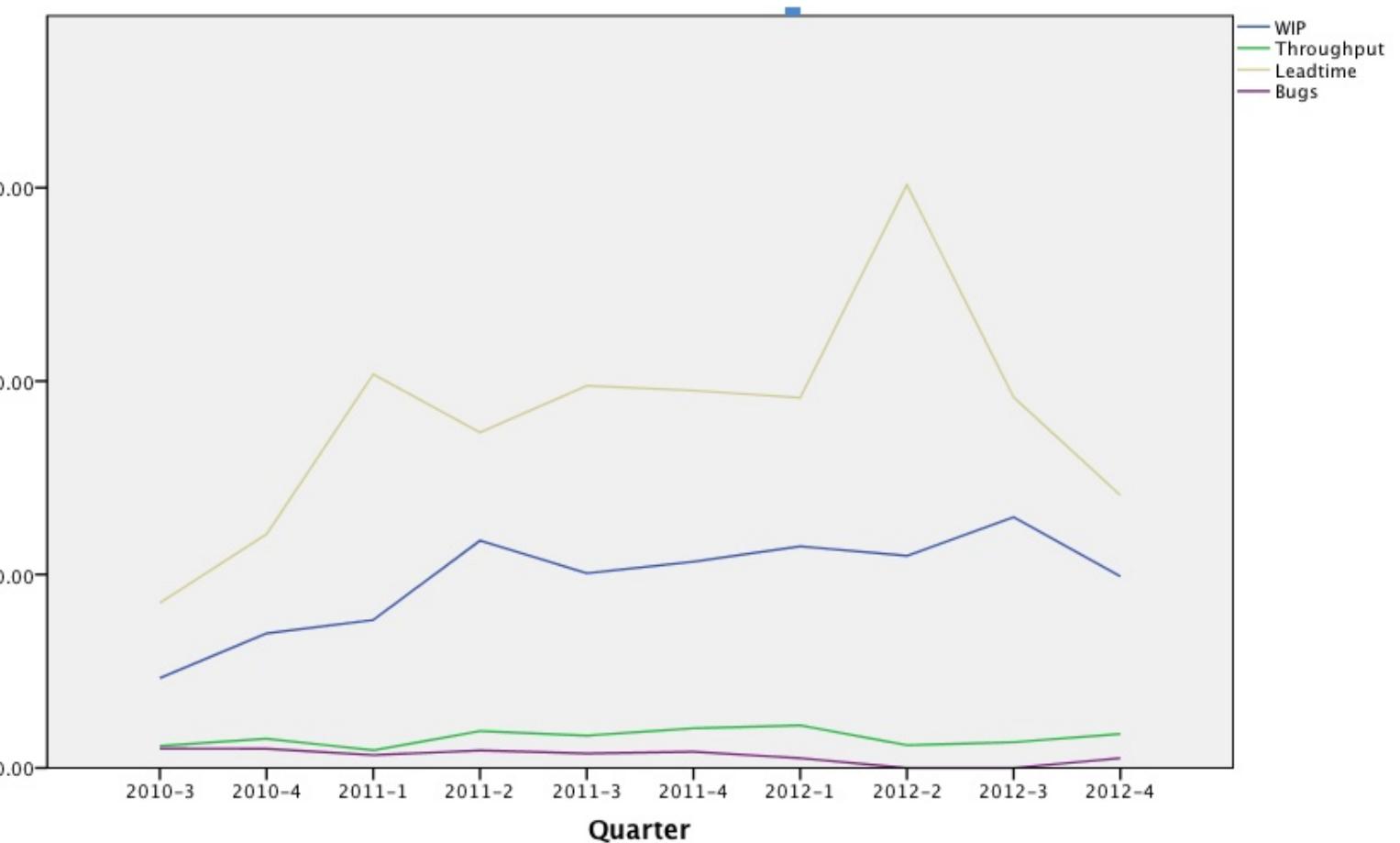


Figure 4.2: WIP, Throughput, Lead time and Bugs for Frontend per quarter

In figure 4.3 the average WIP, Throughput, Lead time and Bugs per quarter for team Krypton are shown

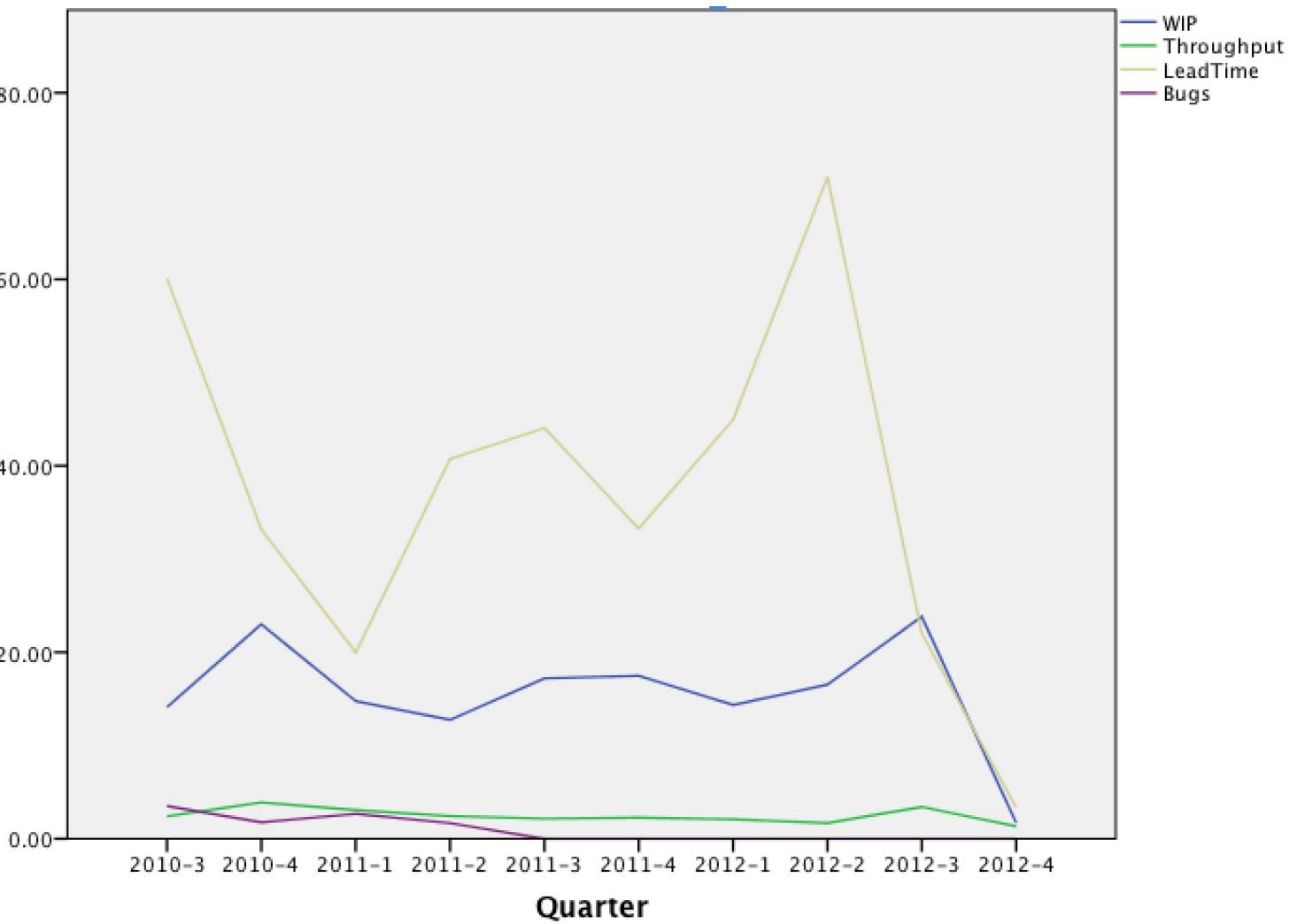
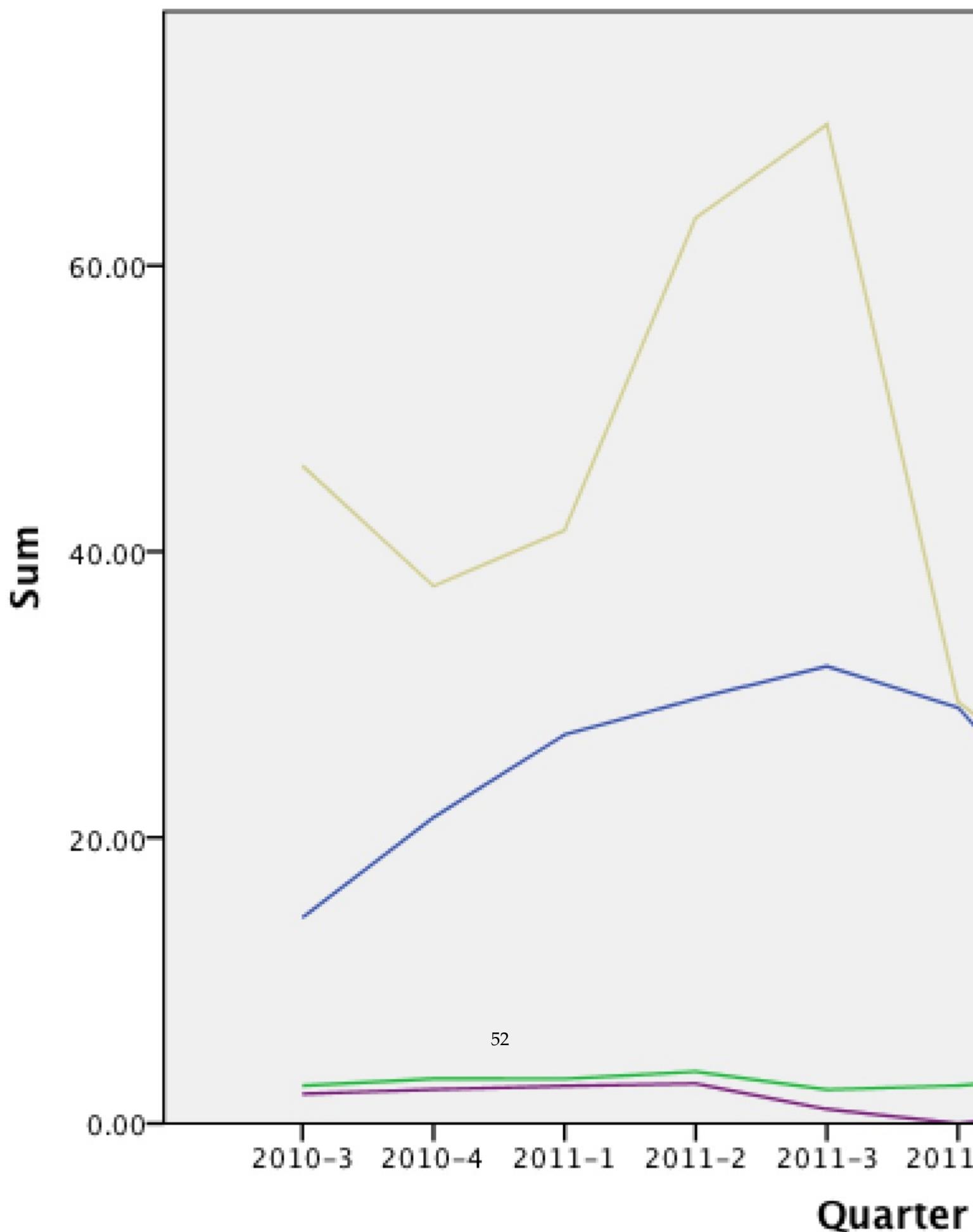


Figure 4.3: WIP, Throughput, Lead time and Bugs for Krypton per quarter

In figure 4.4 the average WIP, Throughput, Lead time and Bugs per quarter for team Neon are shown



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4.4 360

Correlations		Bugs	Throughput	WIP	Churn	TP_Feature	Churn_Bugs	Leadtime_Upon	Churn_Union	Churn_feature	TP_bugs	Average_Days_Backlog	Bugs_churn_average	Average_ft_churn	Precent_bugs_finished	Churn_Average	
Bugs	Pearson Correlation	1	.945**	.753*	.992**	-.328	.813**	-.240	0.433	.730*	.984**	0.377	0.581	0.119	0.543	0.622	
	Sig. (2-tailed)		0	0.012	0.012	0.354	0.004	0.505	0.212	0.026	0	0.282	0.078	0.744	0.104	0.055	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Throughput	Pearson Correlation	.945**	1	.816**	.962**	-.340	.776**	-.177	0.399	.736*	.940**	0.274	0.626	0.030	0.420	.638*	
	Sig. (2-tailed)		0	0.004	0	0.336	0.008	0.624	0.253	0.024	0	0.443	0.053	0.934	0.227	0.047	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
WIP	Pearson Correlation	.753*	.816**	1	.738*	-.441	.708*	-.336	0.365	.0467	.800**	0.397	0.430	0.243	0.495	0.478	
	Sig. (2-tailed)		0.012	0.004		0.015	0.202	0.022	0.342	0.299	0.205	0.005	0.255	0.215	0.498	0.146	0.163
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Churn	Pearson Correlation	.992**	.962**	.738*	1	-.317	.804**	-.203	0.372	.741*	.974**	0.349	0.581	0.087	0.475	0.617	
	Sig. (2-tailed)		0	0	0.015	0.372	0.005	0.574	0.289	0.022	0	0.323	0.078	0.811	0.165	0.057	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
TP_Feature	Pearson Correlation	-.328	-.340	-.441	-.317	1	-.419	0.376	-.460	-.208	-.360	0.097	-.171	-.295	-.795**	-.223	
	Sig. (2-tailed)		0.354	0.336	0.202	0.372	0.228	0.284	0.181	0.591	0.307	0.789	0.637	0.408	0.006	0.536	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Churn_Bugs	Pearson Correlation	.813**	.776**	.708*	.804*	-.419	1	-.126	0.516	.0597	.884**	-.700*	0.560	0.523	0.551	.865*	
	Sig. (2-tailed)		0.004	0.008	0.022	0.005	0.228	0.729	0.127	0.090	0.001	0.024	0.092	0.121	0.099	0.036	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Leadtime_Upon	Pearson Correlation	-.240	-.177	-.336	-.203	0.376	-.126	1	-.060	0.138	-.280	-.012	-.003	0.158	-.0598	0.059	
	Sig. (2-tailed)		0.505	0.624	0.342	0.574	0.284	0.729	0.868	0.724	0.433	0.974	0.994	0.663	0.068	0.871	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Churn_Union	Pearson Correlation	0.433	0.399	0.365	0.372	-.460	0.516	-.060	1	0.202	0.466	0.231	0.146	0.410	.634*	0.198	
	Sig. (2-tailed)		0.212	0.253	0.299	0.289	0.181	0.127	0.868	0.603	0.175	0.522	0.687	0.239	0.049	0.584	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Churn_feature	Pearson Correlation	.730*	.736*	0.467	.741*	-.208	0.597	0.138	0.202	1	.681*	0.073	.924**	0.008	0.362	.947**	
	Sig. (2-tailed)		0.026	0.024	0.205	0.022	0.591	0.090	0.724	0.603	0.044	0.852	0	0.984	0.338	0	
N		9	9	9	9	9	9	9	9	9	9	9	9	9	9	9	
TP_bugs	Pearson Correlation	.984**	.940**	.800**	.974**	-.360	.384**	-.280	0.466	.681*	1	0.494	0.574	0.192	0.568	0.620	
	Sig. (2-tailed)		0	0	0.005	0	0.307	0.001	0.433	0.175	0.044	1	0.146	0.083	0.596	0.087	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Average_Days_Backlog	Pearson Correlation	0.377	0.274	0.397	0.349	0.097	.700*	-.012	0.231	0.073	0.494	1	0.101	0.595	0.130	0.217	
	Sig. (2-tailed)		0.282	0.443	0.255	0.323	0.789	0.024	0.974	0.522	0.852	0.146	0.782	0.070	0.721	0.548	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Bugs_churn_average	Pearson Correlation	0.581	0.626	0.430	0.581	-.171	0.560	-.003	0.146	.924**	0.574	0.101	1	-.127	0.322	.973**	
	Sig. (2-tailed)		0.078	0.053	0.215	0.078	0.637	0.092	0.994	0.687	0	0.083	0.782	0.726	0.365	0	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Average_ft_churn	Pearson Correlation	0.119	0.030	0.243	0.087	-.295	0.523	0.158	0.410	0.008	0.192	0.595	-.127	1	0.291	0.088	
	Sig. (2-tailed)		0.744	0.934	0.498	0.811	0.408	0.121	0.663	0.239	0.984	0.596	0.070	0.726	0.415	0.808	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Precent_bugs_finished	Pearson Correlation	0.543	0.420	0.495	0.475	-.795**	0.551	-.598	0.634*	0.362	0.568	0.130	0.322	0.291	1	0.369	
	Sig. (2-tailed)		0.104	0.227	0.14	0.165	0.006	0.099	0.068	0.049	0.338	0.087	0.721	0.365	0.415	0.294	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	
Churn_Average	Pearson Correlation	0.622	.638*	0.478	0.617	-.223	.665*	0.059	0.198	.947**	0.620	0.217	.973**	0.088	0.369	1	
	Sig. (2-tailed)		0.055	0.047	0.163	0.057	0.536	0.036	0.871	0.584	0	0.056	0.548	0	0.808	0.294	
N		10	10	10	10	10	10	10	10	9	10	10	10	10	10	10	

** Correlation is significant at the 0.01 level (2-tailed).

* Correlation is significant at the 0.05 level (2-tailed).

Table 4.22: 360 - Correlation

Team	Quarter		Churn	Leadtime
360	2010-3	N	11	11
		Mean	17	3.55
		Median	11	3
		Std. Deviation	18.93	2.50
		Maximum	60	8
		Minimum	0	1
	2010-4	N	3	3
		Mean	0.67	1.67
		Median	1	2
		Std. Deviation	0.57	0.57
		Maximum	1	2
		Minimum	0	1
		N	136	136
	2011-2	Mean	14.4	2.7
		Median	3	2
		Std. Deviation	21.51	1.85
		Maximum	83	8
		Minimum	0	1
		N	48	48
	2011-3	Mean	1.67	3.06
		Median	0	2.5
		Std. Deviation	7.71	2.13
		Maximum	50	8
		Minimum	0	1
		N	156	156
	2011-4	Mean	14.81	2.25
		Median	6.5	2
		Std. Deviation	19.90	1.53
		Maximum	86	8
		Minimum	0	1
		N	157	157
	2012-1	Mean	9.47	2.68
		Median	0	2
		Std. Deviation	19.54	1.758
		Maximum	86	8
		Minimum	0	1
		N	238	238
	2012-2	Mean	10.11	2.78
		Median	0	2
		Std. Deviation	18.50	1.597
		Maximum	83	8
		Minimum	0	1
		N	53	53
	2012-3	Mean	3.02	1.68
		Median	0	1
		Std. Deviation	9.61	1.18
		Maximum	53	7
		Minimum	0	1
		N	26	26
	2012-4	Mean	21.42	2.08
		Median	14.5	2
		Std. Deviation	20.54	1.16
		Maximum	67	5
		Minimum	1	1
		N	902	902
	Total	Mean	10.14	2.7
		Median	0	2
		Std. Deviation	18.55	1.798
		Maximum	86	8
		Minimum	0	1

Table 4.23: Descriptive statistic - Lead-time and churn - 360

Frontend

Correlations			WIP	Throughput	Bugs	Churn	TP_feature	Churn_Bugs	Churn_union	Leadtime_union	Churn_feature	Tp_bugs	Average_Days_Backlog_Bugs	Bugs_Churn_average	Churn_feature_Average	Precent_bugs_fininished	Churn_Average
WIP	Pearson Correlation	1	0.491	0.278	-0.446	0.380	-0.446	0.316	0.096	0.362	0.508	-0.178	0.257	0.337	0.367	0.223	
	Sig. (2-tailed)		0.150	0.436	0.196	0.279	0.196	0.373	0.791	0.304	0.134	0.622	0.473	0.341	0.297	0.537	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Throughput	Pearson Correlation	0.491	1	.773**	-0.036	.741*	0.217	-0.494	-0.539	-0.433	.758*	0.055	0.259	-0.609	.745*	-0.575	
	Sig. (2-tailed)		0.150	0.009	0.921	0.014	0.548	0.147	0.108	0.212	0.011	0.879	0.470	0.062	0.013	0.082	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Bugs	Pearson Correlation	0.278	.773**	1	0.259	0.534	0.398	-0.555	-0.581	-0.478	0.271	0.005	0.296	-0.571	0.498	-0.624	
	Sig. (2-tailed)		0.436	0.009	0.471	0.111	0.254	0.096	0.078	0.162	0.449	0.988	0.407	0.085	0.143	0.054	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn	Pearson Correlation	-0.446	-0.036	0.259	1	-0.219	.791**	-0.522	-0.449	-0.517	-0.110	-0.174	0.172	-0.485	-0.014	-0.471	
	Sig. (2-tailed)		0.196	0.921	0.471	0.543	0.006	0.122	0.192	0.126	0.762	0.631	0.634	0.155	0.970	0.170	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
TP_feature	Pearson Correlation	0.380	.741*	0.534	-0.219	1	0.214	-0.244	-0.312	-0.026	0.427	-0.256	-0.007	-0.294	.640*	-0.319	
	Sig. (2-tailed)		0.279	0.014	0.111	0.543	0.553	0.497	0.381	0.943	0.219	0.475	0.984	0.409	0.046	0.369	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_Bugs	Pearson Correlation	-0.446	0.217	0.398	.791**	0.214	1	-.707*	-.771**	-0.460	0.040	-0.378	-0.035	-0.590	0.388	-.689*	
	Sig. (2-tailed)		0.196	0.548	0.254	0.006	0.553	0.022	0.009	0.181	0.912	0.282	0.924	0.073	0.267	0.027	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_union	Pearson Correlation	0.316	-0.494	-0.555	-0.522	-0.244	-.707*	1	.704*	.844**	-0.368	-0.139	0.152	.841**	-.312	.981**	
	Sig. (2-tailed)		0.373	0.147	0.096	0.122	0.497	0.022	0.023	0.002	0.296	0.702	0.674	0.002	0.379	0	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Leadtime_union	Pearson Correlation	0.096	-.539	-0.581	-0.449	-0.312	-.771**	.704*	1	0.492	-0.301	0.327	-0.227	0.631	-.697*	.766**	
	Sig. (2-tailed)		0.791	0.108	0.078	0.192	0.381	0.009	0.023	0.148	0.398	0.357	0.528	0.050	0.025	0.010	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_feature	Pearson Correlation	0.362	-0.433	-0.478	-0.517	-0.026	-0.460	.844**	0.492	1	-0.314	-0.358	-0.201	.921**	-0.172	.842**	
	Sig. (2-tailed)		0.304	0.212	0.162	0.126	0.943	0.181	0.002	0.148	0.377	0.310	0.579	0	0.635	0.002	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Tp_bugs	Pearson Correlation	0.508	.758*	0.271	-0.110	0.427	0.040	-0.368	-0.301	-0.314	1	0.096	0.034	-0.393	0.622	-0.406	
	Sig. (2-tailed)		0.134	0.011	0.449	0.762	0.219	0.912	0.296	0.398	0.377	0.792	0.926	0.261	0.055	0.245	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Average_Days_Backlog_Bugs	Pearson Correlation	-0.178	0.055	0.005	-0.174	-0.256	-0.378	-0.139	0.327	-0.358	0.096	1	-0.267	-0.272	-0.489	-0.035	
	Sig. (2-tailed)		0.622	0.879	0.988	0.631	0.475	0.282	0.702	0.357	0.310	0.792	0.456	0.446	0.151	0.925	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Bugs_Churn_average	Pearson Correlation	0.257	0.259	0.296	0.172	-0.007	-0.035	0.152	-0.227	-0.201	0.034	-0.267	1	-0.197	0.371	0.016	
	Sig. (2-tailed)		0.473	0.470	0.407	0.634	0.984	0.924	0.674	0.528	0.579	0.926	0.456	0.585	0.291	0.966	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_feature_Average	Pearson Correlation	0.337	-0.609	-0.571	-0.485	-0.294	-0.590	.841**	0.631	.921**	-0.393	-0.272	-0.197	1	-0.389	.846**	
	Sig. (2-tailed)		0.341	0.062	0.085	0.155	0.409	0.073	0.002	0.050	0	0.261	0.446	0.585	0.267	0.002	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Precent_bugs_finished	Pearson Correlation	0.367	.745*	0.498	-0.014	.640*	0.388	-0.312	-.697*	-0.172	0.622	-0.489	0.371	-0.389	1	-0.445	
	Sig. (2-tailed)		0.297	0.013	0.143	0.970	0.046	0.267	0.379	0.025	0.635	0.055	0.151	0.291	0.267	0.197	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_Average	Pearson Correlation	0.223	-0.575	-0.624	-0.471	-0.319	-.689*	.981**	.766**	.842**	-0.406	-0.035	0.016	.846**	-0.445	1	
	Sig. (2-tailed)		0.537	0.082	0.054	0.170	0.369	0.027	0	0.010	0.002	0.245	0.925	0.966	0.002	0.197	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	

** Correlation is significant at the 0.01 level (2-tailed).

* Correlation is significant at the 0.05 level (2-tailed).

Table 4.24: Frontend Correlation

Team	Quarter		Leadtime	Churn
Frontend	2010-3	N	58	58
		Mean	7.43	127.93
		Median	6	53.5
		Std. Deviation	5.567	187.011
		Maximum	22	1072
		Minimum	1	0
	2010-4	N	130	130
		Mean	10.32	158.72
		Median	4	14
		Std. Deviation	15.30	444.805
		Maximum	78	4308
		Minimum	1	0
	2011-1	N	95	95
		Mean	28.01	566.73
		Median	11	47
		Std. Deviation	52.183	3696.125
		Maximum	434	35917
		Minimum	1	0
	2011-2	N	132	132
		Mean	12.17	496.8
		Median	5	19
		Std. Deviation	23.548	3240.172
		Maximum	184	35956
		Minimum	1	0
	2011-3	N	146	146
		Mean	10.66	174.07
		Median	6	14
		Std. Deviation	12.577	655.578
		Maximum	79	5650
		Minimum	1	0
	2011-4	N	167	167
		Mean	9.85	110.98
		Median	7	3
		Std. Deviation	10.9	480.32
		Maximum	62	5672
		Minimum	1	0
	2012-1	N	188	188
		Mean	13.9	118.64
		Median	8	5.5
		Std. Deviation	21.92	568.03
		Maximum	150	7565
		Minimum	1	0
	2012-2	N	77	77
		Mean	22.53	849.42
		Median	17	15
		Std. Deviation	25.52	3575.50
		Maximum	106	27170
		Minimum	1	0
	2012-3	N	93	93
		Mean	13.76	458.92
		Median	8	25
		Std. Deviation	18.023	1272.32
		Maximum	107	8600
		Minimum	1	0
	2012-4	N	82	82
		Mean	8.8007	428.67
		Median	4	54
		Std. Deviation	14.978	1178.471
		Maximum	83	9368
		Minimum	1	0
Total		N	1168	1168
		Mean	13.35	305.62
		Median	7	15
		Std. Deviation	23.15	1883.16
		Maximum	434	35956
		Minimum	1	0

Table 4.25: Descriptive statistic - Lead-time and churn - Frontend

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Frontend	2010-3	14	2.79	2	2.91	12	1
	2010-4	39	2	1	1.53	8	1
	2011-1	20	1.85	1	1.226	5	1
	2011-2	29	2.86	2	3.21	16	1
	2011-3	37	2.41	2	1.93	10	1
	2011-4	35	2.89	1	5.02	30	1
	2012-1	39	3.56	2	3.90	23	1
	2012-2	31	1.81	1	1.4	7	1
	2012-3	24	2.54	2	1.79	7	1
	2012-4	14	2.86	2	1.79	7	1
Total		282	2.56	2	2.895	30	1

Table 4.26: Frontend - Descriptive statistic - Bugs

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Frontend	2010-3	8	1.25	1	0.46	2	1
	2010-4	35	1.63	1	0.877	4	1
	2011-1	32	1.53	1	0.67	3	1
	2011-2	31	1.74	2	0.89	4	1
	2011-3	31	1.97	2	1.25	6	1
	2011-4	28	1.93	1	1.94	11	1
	2012-1	22	2.18	1	2.01	7	1
	2012-2	23	1.48	1	1.123	6	1
	2012-3	21	1.48	1	0.873	4	1
	2012-4	19	2.47	2	1.679	8	1
Total		250	1.78	1	1.294	11	1

Table 4.27: Frontend - Descriptive statistic - TP feature

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Frontend	2010-3	15	2	2	1.363	6	1
	2010-4	41	1.98	2	1.387	8	1
	2011-1	18	1.33	1	0.48	2	1
	2011-2	35	2.50	2	2.161	10	1
	2011-3	38	2.18	2	1.60	8	1
	2011-4	40	2.6	2	2.52	14	1
	2012-1	41	3.44	3	2.61	10	1
	2012-2	29	1.9	1	1.472	7	1
	2012-3	30	2.13	1.5	1.69	8	1
	2012-4	25	2.12	1	2.10	10	1
Total		312	2.31	2	1.982	14	1

Table 4.28: Frontend - Descriptive statistic - TP bugs

4.5 Krypton

Correlations		WIP	Throughput	Bugs	Churn	TP_feature	Churn_Bugs	Churn_union	Leadtime_union	Churn_feature	Tp_bugs	Average_Days	Backlog_Bugs	Bugs_Churn_average	Churn_feature_Average	Percent_bugs_finished	Churn
WIP	Pearson Correlation	1	0.491	0.278	-0.446	0.380	-0.446	0.316	0.096	0.362	0.508	-0.178	0.257	0.337	0.367	0.223	
	Sig. (2-tailed)		0.150	0.436	0.196	0.279	0.196	0.373	0.791	0.304	0.134	0.622	0.473	0.341	0.297	0.535	
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Throughput	Pearson Correlation	0.491	1	.773**	-0.036	.741*	0.217	-0.494	-0.539	-0.433	.758*	0.055	0.259	-0.609	.745*	-0.57
	Sig. (2-tailed)		0.150	0.009	0.921	0.014	0.548	0.147	0.108	0.212	0.011	0.879	0.470	0.062	0.013	0.082	0.10
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Bugs	Pearson Correlation	0.278	.773**	1	0.259	0.534	0.398	-0.555	-0.581	-0.478	0.271	0.005	0.296	-0.571	0.498	-0.62
	Sig. (2-tailed)		0.436	0.009	0.471	0.111	0.254	0.096	0.078	0.162	0.449	0.988	0.407	0.085	0.143	0.054	0.10
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Churn	Pearson Correlation	-0.446	-0.036	0.259	1	-0.219	.791**	-0.522	-0.449	-0.517	-0.110	-0.174	0.172	-0.485	-0.014	-0.47
	Sig. (2-tailed)		0.196	0.921	0.471	0.543	0.006	0.122	0.192	0.126	0.762	0.631	0.634	0.155	0.970	0.170	0.10
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
TP_feature	Pearson Correlation	0.380	.741*	0.534	-0.219	1	0.214	-0.244	-0.312	-0.026	0.427	-0.256	-0.007	-0.294	-0.640*	-0.31	
	Sig. (2-tailed)		0.279	0.014	0.111	0.543	0.553	0.497	0.381	0.943	0.219	0.475	0.984	0.409	0.046	0.369	0.10
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Churn_Bugs	Pearson Correlation	-0.446	0.217	0.398	.791**	0.214	1	-.707*	-.771**	-0.460	0.040	-0.378	-0.035	-0.590	0.388	-0.689
	Sig. (2-tailed)		0.196	0.548	0.254	0.006	0.553	0.022	0.009	0.181	0.912	0.282	0.924	0.073	0.267	0.027	0.10
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Churn_union	Pearson Correlation	0.316	-0.494	-0.555	-0.522	-0.244	-.707*	1	.704*	.844**	-0.368	-0.139	0.152	.841**	-0.312	.981*
	Sig. (2-tailed)		0.373	0.147	0.096	0.122	0.497	0.022	0.023	0.002	0.296	0.702	0.674	0.002	0.379	0	0
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Leadtime_union	Pearson Correlation	0.096	-0.539	-0.581	-0.449	-0.312	-.771**	.704*	1	0.492	-0.301	0.327	-0.227	0.631	-.697*	.766*
	Sig. (2-tailed)		0.791	0.108	0.078	0.192	0.381	0.009	0.023	0.148	0.398	0.357	0.528	0.050	0.025	0.010	0.010
Churn_feature	Pearson Correlation	0.362	-0.433	-.0478	-0.517	-0.026	-0.460	.844**	1	0.492	1	-0.314	-0.358	-0.201	.921**	-0.172	.842*
	Sig. (2-tailed)		0.304	0.212	0.162	0.126	0.943	0.181	0.002	0.148	1	0.377	0.310	0.579	0	0.635	0.002
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Tp_bugs	Pearson Correlation	0.508	.758*	.271	-0.110	0.427	0.040	-0.368	-0.301	-0.314	1	0.096	0.034	-0.393	0.622	-0.40
	Sig. (2-tailed)		0.134	0.011	0.449	0.762	0.219	0.912	0.296	0.398	0.377	0.792	0.926	0.261	0.055	0.245	0.10
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Average_Days_Backlog_Bugs	Pearson Correlation	-0.178	0.055	0.005	-.174	-0.256	-.378	-0.139	0.327	-0.358	0.096	1	-0.267	-0.272	-0.489	-0.03
	Sig. (2-tailed)		0.622	0.879	0.988	0.631	0.475	0.282	0.702	0.357	0.310	0.792	0.456	0.446	0.151	0.925	0.10
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Bugs_Churn_average	Pearson Correlation	0.257	0.259	0.296	0.172	-.007	-0.035	0.152	-0.227	-0.201	0.034	-0.267	1	-0.197	0.371	0.016
	Sig. (2-tailed)		0.473	0.470	0.407	0.634	0.984	0.924	0.674	0.528	0.579	0.926	0.456	0.585	0.291	0.966	0.002
Churn_feature_Average	Pearson Correlation	0.337	-0.609	-.571	-0.485	-0.294	-.590	.841**	0.631	.921**	-0.393	-0.272	-0.197	1	-0.389	.846*	
	Sig. (2-tailed)		0.341	0.062	0.085	0.155	0.409	0.073	0.002	0.050	0	0.261	0.446	0.585	0.267	0.002	0.10
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Precent_bugs_finished	Pearson Correlation	0.367	.745*	0.498	-.014	.640*	0.388	-.312	-.697*	-0.172	0.622	-0.489	0.371	-0.389	1	-.044
	Sig. (2-tailed)		0.297	0.013	0.143	0.970	0.046	0.267	0.379	0.025	0.635	0.055	0.151	0.291	0.267	0.197	0.197
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
	Churn_Average	Pearson Correlation	0.223	-.0575	-.624	-.471	-.319	-.689*	.981**	.766**	.842**	-0.406	-0.035	0.016	.846**	-0.445	1
	Sig. (2-tailed)		0.537	0.082	0.054	0.170	0.369	0.027	0	0.010	0.002	0.245	0.925	0.966	0.002	0.197	1
	N	10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	

*Correlation is significant at the 0.05 level (2-tailed).

Table 4.29: Krypton - correlation

Team	Quarter		Leadtime	Churn
Krypton	2010-3	N	46	46
		Mean	7.89	159.38
		Median	3.5	26.5
		Std. Deviation	11.651	206.584
		Maximum	71	647
		Minimum	1	0
	2010-4	N	139	139
		Mean	7.38	97.7
		Median	3	28
		Std. Deviation	12.88	145.84
		Maximum	85	700
		Minimum	1	0
	2011-1	N	153	153
		Mean	5.83	71.38
		Median	2	17
		Std. Deviation	7.383	130.40
		Maximum	32	726
		Minimum	1	0
	2011-2	N	51	51
		Mean	14.37	110.45
		Median	9	31
		Std. Deviation	16.09	165.053
		Maximum	90	719
		Minimum	1	0
	2011-3	N	49	49
		Mean	13.71	119.43
		Median	8	27
		Std. Deviation	15.97	159.39
		Maximum	88	604
		Minimum	1	1
	2011-4	N	66	66
		Mean	10.5	112.36
		Median	6	36
		Std. Deviation	11.975	164.935
		Maximum	57	675
		Minimum	1	1
	2012-1	N	45	45
		Mean	23.73	99.09
		Median	8	23
		Std. Deviation	40.47	149.02
		Maximum	180	636
		Minimum	1	1
	2012-2	N	45	45
		Mean	23	109.07
		Median	7	47
		Std. Deviation	32.08	131.58
		Maximum	135	536
		Minimum	1	1
	2012-3	N	74	74
		Mean	6.92	82.39
		Median	3	37
		Std. Deviation	9.26	122.637
		Maximum	35	713
		Minimum	1	1
	2012-4	N	2	2
		Mean	1.5	44
		Median	1.5	44
		Std. Deviation	0.70	1.41
		Maximum	2	45
		Minimum	1	43
	Total	N	683	683
		Mean	10.55	98.6
		Median	4	29
		Std. Deviation	18.1	149.01
		Maximum	180	726
		Minimum	1	0

Table 4.30: Descriptive statistic - Lead-time and churn - Krypton

Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
2010-3	14.00	3.07	2.00	3.47	14.00	1.00
2010-4	71.00	2.73	2.00	2.05	11.00	1.00
2011-1	67.00	2.55	2.00	1.96	13.00	1.00
2011-2	38.00	1.97	2.00	1.03	4.00	1.00
2011-3	43.00	1.65	1.00	0.92	4.00	1.00
2011-4	53.00	1.83	1.00	1.31	6.00	1.00
2012-1	40.00	1.90	1.00	1.46	7.00	1.00
2012-2	27.00	2.07	2.00	1.36	6.00	1.00
2012-3	40.00	2.70	2.00	2.43	13.00	1.00
2012-4	3.00	1.00	1.00	0	1.00	1.00
Total	396.00	2.26	2.00	1.82	14.00	1.00

Table 4.31: Frontend - Descriptive statistic - Throughput

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Krypton	2010-3	25	14.12	14	5.093	27	9
	2010-4	92	23	23.5	7.819	48	12
	2011-1	90	14.76	13.5	5	26	6
	2011-2	91	12.75	12	4.64	22	3
	2011-3	92	17.2	16	6.476	32	8
	2011-4	92	17.47	17	6.32	30	4
	2012-1	91	14.35	15	3.911	24	7
	2012-2	91	16.53	14	6.03	34	10
	2012-3	92	23.83	20	10.387	43	7
	2012-4	67	1.69	0	3.016	11	0
Total		823	16.11	15	8.43	48	0

Table 4.32: Krypton - Descriptive statistic WIP

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Krypton	2010-3	17	1.76	2	0.83	3	1
	2010-4	28	1.71	1.5	0.85	4	1
	2011-1	39	3.1	2	2.51	12	1
	2011-2	26	2.04	2	1.28	5	1
	2011-3	25	1.64	1	1.14	5	1
	2011-4	26	2	2	1.35	7	1
	2012-1	26	1.85	1	1.617	9	1
	2012-2	19	2.52	2	2.48	11	1
	2012-3	31	2.16	2	1.44	6	1
	Total	241	2.12	2	1.69	12	1
	2010-3	20	2.65	2	3.54	17	1

Table 4.33: Krypton - Descriptive statistic - Bugs

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Krypton	2010-3	10	2.20	1.5	1.619	5	1
	2010-4	48	2.79	2	2.278	11	1
	2011-1	30	1.7	1	0.95	4	1
	2011-2	17	1.65	1	1.115	4	1
	2011-3	20	1.45	1	0.82	4	1
	2011-4	31	1.48	1	0.89	4	1
	2012-1	14	1.71	1	1.139	5	1
	2012-2	11	1.73	1	0.90	3	1
	2012-3	17	1.53	1	0.8	4	1
	2012-4	3	1	1	0	1	1
Total		201	1.9	1	1.48	11	1

Table 4.34: Krypton - Descriptive statistic - TP feature

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Krypton	2010-3	6	3.5	2.5	3.2090	9	1
	2010-4	34	1.76	2	0.89	4	1
	2011-1	45	2.67	2	2.20	13	1
	2011-2	28	1.68	2	0.77	4	1
	2011-3	26	1.62	1	0.752	3	1
	2011-4	27	1.89	1	1.45	6	1
	2012-1	28	1.86	1	1.407	7	1
	2012-2	18	2.06	2	1.552	6	1
	2012-3	28	2.93	3	2.19	9	1
	Total	240	2.13	2	1.66	13	1

Table 4.35: Krypton - Descriptive statistic - TP bugs

4.6 Neon

Correlations			WIP	Throughput	Bugs	Churn	TP_feature	Churn_Bugs	Churn_union	Leadtime_union	Churn_feature	Tp_bugs	Average_Days_Backlog_Bugs	Bugs_Churn_average	Churn_feature_Average	Percent_bugs_fininished	Churn_Average
WIP	Pearson Correlation	1	0.491	0.278	-0.446	0.380	-0.446	0.316	0.096	0.362	0.508	-0.178	0.257	0.337	0.367	0.223	
	Sig. (2-tailed)		0.150	0.436	0.196	0.279	0.196	0.373	0.791	0.304	0.134	0.622	0.473	0.341	0.297	0.0537	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Throughput	Pearson Correlation	0.491	1	.773**	-0.036	.741*	0.217	-0.494	-0.539	-0.433	.758*	0.055	0.259	-0.609	.745*	-0.575	
	Sig. (2-tailed)		0.150	0.009	0.921	0.014	0.548	0.147*	0.108	0.212	0.011	0.879	0.470	0.062	0.013	0.082	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Bugs	Pearson Correlation	0.278	.773**	1	0.259	0.534	0.398	-0.555	-0.581	-0.478	0.271	0.005	0.296	-0.571	0.498	-0.624	
	Sig. (2-tailed)		0.436	0.009	0.471	0.111	0.254	0.096	0.078	0.162	0.449	0.988	0.407	0.085	0.143	0.054	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn	Pearson Correlation	-0.446	-0.036	0.259	1	-0.219	.791**	-0.522	-0.449	-0.517	-0.110	-0.174	0.172	-0.485	-0.014	-0.471	
	Sig. (2-tailed)		0.196	0.921	0.471	0.543	0.006	0.122	0.192	0.126	0.762	0.631	0.634	0.155	0.970	0.170	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
TP_feature	Pearson Correlation	0.380	.741*	0.534	-0.219	1	0.214	-0.244	-0.312	-0.026	0.427	-0.256	-0.007	-0.294	.640*	-0.319	
	Sig. (2-tailed)		0.279	0.014	0.111	0.543	0.553	0.497	0.381	0.943	0.219	0.475	0.984	0.409	0.046	0.369	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_Bugs	Pearson Correlation	-0.446	0.217	0.398	.791**	0.214	1	-0.707*	-0.771**	-0.460	0.040	-0.378	-0.035	-0.590	0.388	-0.689*	
	Sig. (2-tailed)		0.196	0.548	0.254	0.006	0.553	0.022	0.009	0.181	0.912	0.282	0.924	0.073	0.267	0.027	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_union	Pearson Correlation	0.316	-0.494	-0.555	-0.522	-0.244	-0.707*	1	.704*	.844**	-0.368	-0.139	0.152	.841**	-0.312	.981**	
	Sig. (2-tailed)		0.373	0.147	0.096	0.122	0.497	0.022	0.023	0.002	0.296	0.702	0.674	0.002	0.379	0	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Leadtime_union	Pearson Correlation	0.096	-0.539	-0.581	-0.449	-0.312	-0.771**	.704*	1	0.492	-0.301	0.327	-0.227	0.631	-0.697*	.766**	
	Sig. (2-tailed)		0.791	0.108	0.078	0.192	0.381	0.009	0.023	0.148	0.398	0.357	0.528	0.050	0.025	0.010	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_feature	Pearson Correlation	0.362	-0.433	-0.478	-0.517	-0.026	-0.460	.844**	0.492	1	-0.314	-0.358	-0.201	.921**	-0.172	.842**	
	Sig. (2-tailed)		0.304	0.212	0.162	0.126	0.943	0.181	0.002	0.148	0.377	0.310	0.579	0	0.635	0.002	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Tp_bugs	Pearson Correlation	0.508	.758*	0.271	-0.110	0.427	0.040	-0.368	-0.301	-0.314	1	0.096	0.034	-0.393	0.622	-0.406	
	Sig. (2-tailed)		0.134	0.011	0.449	0.762	0.219	0.912	0.296	0.398	0.377	0.792	0.926	0.261	0.055	0.245	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Average_Days_Backlog_Bugs	Pearson Correlation	-0.178	0.055	0.005	-0.174	-0.256	-0.378	-0.139	0.327	-0.358	0.096	1	-0.267	-0.272	-0.489	-0.035	
	Sig. (2-tailed)		0.622	0.879	0.988	0.631	0.475	0.282	0.702	0.357	0.310	0.792	0.456	0.446	0.151	0.925	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Bugs_Churn_average	Pearson Correlation	0.257	0.259	0.296	0.172	-0.007	-0.035	0.152	-0.227	-0.201	0.034	-0.267	1	-0.197	0.371	0.016	
	Sig. (2-tailed)		0.473	0.470	0.407	0.634	0.984	0.924	0.674	0.528	0.579	0.926	0.456	0.585	0.291	0.969	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_feature_Average	Pearson Correlation	0.337	-0.609	-0.571	-0.485	-0.294	-0.590	.841**	0.631	.921**	-0.393	-0.272	-0.197	1	-0.389	.846**	
	Sig. (2-tailed)		0.341	0.062	0.085	0.155	0.409	0.073	0.002	0.050	0	0.261	0.446	0.585	0.267	0.002	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Precent_bugs_fininished	Pearson Correlation	0.367	.745*	0.498	-0.014	.640*	0.388	-0.312	.697*	-0.172	0.622	-0.489	0.371	-0.389	1	-0.445	
	Sig. (2-tailed)		0.297	0.013	0.143	0.970	0.046	0.267	0.379	0.025	0.635	0.055	0.151	0.291	0.267	0.197	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	
Churn_Average	Pearson Correlation	0.223	-0.575	-0.624	-0.471	-0.319	-0.689*	.981**	.766**	.842**	-0.406	-0.035	0.016	.846**	-0.445	1	
	Sig. (2-tailed)		0.537	0.082	0.054	0.170	0.369	0.027	0	0.010	0.002	0.245	0.925	0.966	0.002	0.197	
N		10	10	10	10	10	10	10	10	10	10	10	10	10	10	10	

* Correlation is significant at the 0.05 level (2-tailed).

** Correlation is significant at the 0.01 level (2-tailed).

Table 4.36: Neon - Correlation

Team	Quarter		Churn	Leadtime
Neon	2010-3	N	62	62
		Mean	25.27	7.69
		Median	9	6
		Std. Deviation	39.448	6.88
		Maximum	193	24
		Minimum	0	1
	2010-4	N	125	125
		Mean	33.13	6.66
		Median	15	5
		Std. Deviation	47.46	6.01
		Maximum	214	27
		Minimum	0	1
	2011-1	N	132	132
		Mean	23.81	6.53
		Median	9	5.5
		Std. Deviation	41.22	5.43
		Maximum	301	27
		Minimum	0	1
	2011-2	N	134	134
		Mean	36.1	10.71
		Median	7	6
		Std. Deviation	59.067	35.19
		Maximum	271	408
		Minimum	0	1
	2011-3	N	99	99
		Mean	40.86	8.34
		Median	17	6
		Std. Deviation	50.60	6.61
		Maximum	239	27
		Minimum	1	1
	2011-4	N	98	98
		Mean	63.31	8.5
		Median	38	7
		Std. Deviation	74.35	6.84
		Maximum	294	27
		Minimum	1	1
	2012-1	N	110	110
		Mean	56.75	5.16
		Median	21	4
		Std. Deviation	72.52	4.46
		Maximum	296	22
		Minimum	1	1
	2012-2	N	81	81
		Mean	44.95	5.05
		Median	22	4
		Std. Deviation	54.23	4.43
		Maximum	300	23
		Minimum	1	1
	2012-3	N	174	174
		Mean	66.12	5.78
		Median	33.5	4
		Std. Deviation	81.02	4.35
		Maximum	299	21
		Minimum	1	1
	2012-4	N	120	120
		Mean	75.93000000000007	5.3
		Median	32	4.5
		Std. Deviation	86.49	4.
		Maximum	303	17
		Minimum	1	1
	Total	N	1155	1155
		Mean	47.76	7
		Median	18	5
		Std. Deviation	66.37	13.112
		Maximum	303	408
		Minimum	0	1

Table 4.37: Descriptive statistic - Lead-time and churn - Neon

Team	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Neon	2010-3	10	2	1.5	1.333	5	1
	2010-4	34	1.68	1	1.173	6	1
	2011-1	31	1.71	1	1.03	5	1
	2011-2	6	1.17	1	0.4	2	1
	2011-3	9	1	1	0	1	1
	2011-4	1	1	1	.	1	1
	Total	91	1.62	1	1.06	6	1

Table 4.38: Neon - Descriptive statistic - Throughput

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Neon	2010-3	25	14.4	15	6.19	23	6
	2010-4	92	21.41	20	7.17	41	9
	2011-1	90	27.2	27.5	4.90	38	17
	2011-2	91	29.71	27	14.42	62	12
	2011-3	92	31.98	30	8.90	55	18
	2011-4	92	29.09	29	10.09	45	12
	2012-1	91	19.03	18	4.63	30	7
	2012-2	91	24.34	25	10.282	50	5
	2012-3	92	23.24	21.5	7.89	44	10
	2012-4	92	19.48	22	10.99	45	1
	2013-1	11	2.18	2	0.60	3	1
	Total	859	24.45	24	10.54	62	1

Table 4.39: Neon - Descriptive statistic - WIP

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Neon	2010-4	40	2.45	2	1.694	9	1
	2011-1	47	2.43	2	1.80	8	1
	2011-2	42	3.57	3	2.52	13	1
	2011-3	45	2.47	2	2.33	13	1
	2011-4	48	2.48	2	1.59	7	1
	2012-1	36	3.25	3	3.03	16	1
	2012-2	36	2.19	2	1.48	8	1
	2012-3	44	3.43	2	2.55	10	1
	2012-4	33	3.12	2	2.63	10	1
	2013-1	1	1	1	.	1	1
	Total	404	2.75	2	2.306	17	1

Table 4.40: Neon - Descriptive statistic - Bugs

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Neon	2010-3	10	2	1.5	1.333	5	1
	2010-4	34	1.68	1	1.173	6	1
	2011-1	31	1.71	1	1.038	5	1
	2011-2	11	1.27	1	0.64	3	1
	2011-3	34	1.53	1	0.99	5	1
	2011-4	16	1.5	1	0.73	3	1
	2012-1	23	1.3	1	0.70	4	1
	2012-2	32	1.63	1	0.83	4	1
	2012-3	32	2.02	2	0.93	4	1
	2012-4	40	1.93	2	1.11	6	1
Total		263	1.69	1	0.997	6	1

Table 4.41: Neon - Descriptive statistic - TP feature

TeamName	Quarter	N	Mean	Median	Std. Deviation	Maximum	Minimum
Neon	2010-3	10	2	1.5	1.333	5	1
	2010-4	34	1.68	1	1.173	6	1
	2011-1	31	1.71	1	1.038	5	1
	2011-2	11	1.27	1	0.64	3	1
	2011-3	34	1.53	1	0.99	5	1
	2011-4	16	1.5	1	0.73	3	1
	2012-1	23	1.3	1	0.70	4	1
	2012-2	32	1.63	1	0.83	4	1
	2012-3	32	2.02	2	0.93	4	1
	2012-4	40	1.93	2	1.11	6	1
Total		263	1.69	1	0.997	6	1

Table 4.42: Neon - Descriptive statistic - TP bugs

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