Volunteering with DataKind: Predicting Foster Care Case Complexity

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Prepared for Orlando Machine Learning and Data Science

About Me

- Kevin Kho
- BS/MS Civil Engineering from University of Illinois at Urbana-Champaign
- 1 year research with USGS
- 2 years as a data scientist with Itron
- 1 year as a data science mentor with Thinkful
- How I ended up volunteering

Acknowledgements

- DataKind
- CBCCF Matt Baker
- Microsoft Cities
- Volunteer Team:
 - Anasuya Das, Alice Feng, Margaret Furr, Juan Bages, Sebastian Ouellet, Matt Robinson, Sebastien Ouellet, Sajala Shukla, Min Li
- ** Most of the slides for this presentation were the work of Alice Feng

Community Based Care of Central Florida

- Non-profit organization responsible for foster care, adoption, and diversion
- Since 2004, CBCCF has facilitated 1000 adoptions, reduced child removal rate by 25%, successfully reunited 69% of families, and doubled the number of foster homes
- Also have programs for minors to get livelihood training

JOB STRESS









Caseload Overload

Job Trauma

EFT THEIR JOB





\$54K TO REPLACE A SINGLE WORKER



· Causing ······

KIDS to WAIT

- · Child's case must be re-assigned.
- New caseworker must have 102 hours of core training prior to carrying full caseload.
- · Must then re-establish child & family relationships.



... meanwhile, the child waits.

Community Based Care of Central Florida

- In a case study conducted by Milwaukee County, children who have one caseworker achieve permanency 74.5 percent of the time. With two caseworkers the chance of permanency decreases to 17.5 percent
- Community Based Care of Central Florida, where training new caseworkers costs between 25%-40% of their annual salary

Community Based Care of Central Florida

- Met with DataKind to scope projects to increase retention of caseworkers
- 2 Projects came out:
 - a. Planning Optimization Tool
 - b. Measurement of Case Complexity

Expected Timeline

October 2017	Team Formation
Early December 2017	Project Start
Late January 2018	EDA
March 2018	Initial Deliverables
May 2018	Final Deliverables
June 2018	Handoff

Actual Timeline

October 2017	Team Formation
Early December 2017	Project Start
February 2018	EDA
April 2018	Initial Deliverables
July 2018	Final Deliverables
August 2018 (2 months)	Handoff

Project Structure

- Project Manager
- Data Ambassador
- 2 Tech Leads
 - o 3 Members Each

Relationship of Two Projects

- Geospatial Component
- Cases with siblings
- Where does the caseworker already travel to?

Scheduling Optimization Tool

- Case workers are the intended user
- The tool helps case workers decide how to sequence tasks
- Users log into the tool once per week
- Solutions should be flexible
- The system may be able to leverage information about activity on a particular case to identify upcoming due dates

Scheduling Optimization Tool

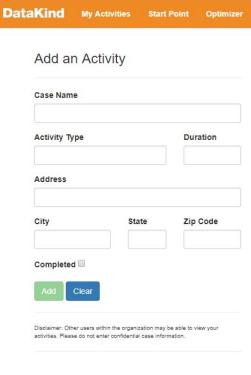
- Angular
 - Computes an optimal schedule for a case worker, given the case worker's activities and constraints
 - Allows users to modify a schedule after it has been optimized
 - Pulls in individual user information based on an id in the URL
 - Modified baseline Angular configurations to ensure that it is compatible with Internet Explorer

Scheduling Optimization Tool

- Built a backend application using Python and MySQL
 - Delivered a unit tested Python package that includes web services, and interface with the COIN-OR integer programming solver, and utilities to help manage the application
 - MySQL database with an automated build process

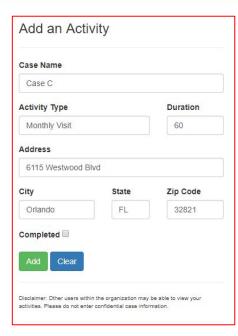
Optimization App Features

User Interface





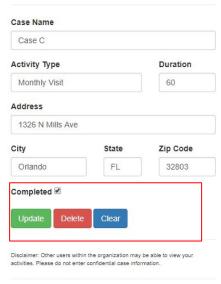
Adding Activities



Expected Duration: 60 minutes	Completed: No			
Address: 1326 N Mills Ave	City: Orlando	State: FL	Zip Code: 32803	
Personal Birthday Party				
Expected Duration: 120 minutes	Completed: No			
Address: 8625 International Dr	City: Orlando	State: FL	Zip Code : 32819	
Case A Monthly Visit				
Expected Duration: 60 minutes	Completed: No			
Address: 2000 Premier Row	City: Orlando	State: FL	Zip Code: 32809	
Case A Court Visit				
Expected Duration: 120 minutes	Completed: No			
Address: 22 E Pine St	City: Orlando	State: FL	Zip Code: 32801	
CBCCF Team Building				
Expected Duration: 50 minutes	Completed: No			
Address: 5803 Precision Dr	City: Orlando	State: FL	Zip Code: 32819	

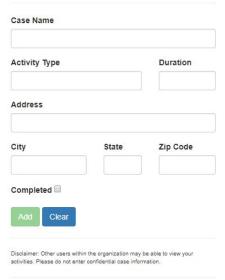
Marking Completed Activities

Add an Activity



Expected Duration: 60 minutes	Completed: No		
Address: 1326 N Mills Ave	City: Orlando	State: FL	Zip Code: 32803
Personal Birthday Party			
Expected Duration: 120 minutes	Completed: No		
Address: 8625 International Dr	City: Orlando	State: FL	Zip Code: 32819
Case A Monthly Visit			
Expected Duration: 60 minutes	Completed: No		
Address: 2000 Premier Row	City: Orlando	State: FL	Zip Code: 32809
Case A Court Visit			
Expected Duration: 120 minutes	Completed: No		
Address: 22 E Pine St	City: Orlando	State: FL	Zip Code: 32801
CBCCF Team Building			
Expected Duration: 50 minutes	Completed: No		
Address: 5803 Precision Dr	City: Orlando	State: FL	Zip Code: 32819
Case C Monthly Visit			
Expected Duration: 60 minutes	Completed: No		
Address: 6115 Westwood Blvd	City: Orlando	State: FI	Zip Code: 32821

Add an Activity



Expected Duration: 60 minutes	Completed: Yes		
Address: 1326 N Mills Ave	City: Orlando	State: FL	Zip Code: 32803
Personal Birthday Party			
Expected Duration: 120 minutes	Completed: No		
Address: 8625 International Dr	City: Orlando	State: FL	Zip Code: 32819
Case A Monthly Visit			
Expected Duration: 60 minutes	Completed: No		
Address: 2000 Premier Row	City: Orlando	State: FL	Zip Code: 32809
Case A Court Visit			
Expected Duration: 120 minutes	Completed: No		
Address: 22 E Pine St	City: Orlando	State: FL	Zip Code: 32801
CBCCF Team Building			
Expected Duration: 50 minutes	Completed: No		
Address: 5803 Precision Dr	City: Orlando	State: FL	Zip Code: 32819
Case C Monthly Visit			
Expected Duration: 60 minutes	Completed: No		
Address: 6115 Westwood Blvd	City: Orlando	State: FL	Zip Code: 32821

Setting Starting/ Ending Points

Start Point			End Point		
Current Start: 4001 Pelee St, Orlando, FL, 32817			Current End: 4001 Pelee St, Orlando, FL, 32817		
Address			Address		
City	State	Zip Code	City	State	Zip Code

DataKind My Activities Start Point Optimizer

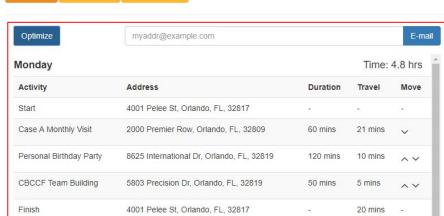
Optimizer

Map

Schedule Optimizer

Schedule





Tuesday			Time:	5.6 hrs
Activity	Address	Duration	Travel	Move
Start	4001 Pelee St, Orlando, FL, 32817	127	2	2
Case A Court Visit	22 E Pine St, Orlando, FL, 32801	120 mins	13 mins	^ ~

DataKind My Activities Start Point Optimizer

Ordering Activities

Мар

Schedule Optimizer

Start

Case A Court Visit



Optimize	myaddr@example.com			E-ma
Monday			Time:	4.9 hrs
Activity	Address	Duration	Travel	Move
Start	4001 Pelee St, Orlando, FL, 32817	2	12	121
CBCCF Team Building	5803 Precision Dr, Orlando, FL, 32819	50 mins	20 mins	~
Case A Monthly Visit	2000 Premier Row, Orlando, FL, 32809	60 mins	8 mins	^ ~
Personal Birthday Party	8625 International Dr, Orlando, FL, 32819	120 mins	10 mins	^ ~
Finish	4001 Pelee St, Orlando, FL, 32817	-	23 mins	(5)
Tuesday			Time:	5.6 hrs
Activity	Address	Duration	Travel	Move

4001 Pelee St. Orlando, FL, 32817

22 E Pine St, Orlando, FL, 32801

13 mins

120 mins

Map

Schedule Optimizer

E-mail

Move

^ V

~ V

Move

^ V

Time: 4.9 hrs

Travel

20 mins

8 mins

10 mins

23 mins

Travel

13 mins

Time: 5.6 hrs

Duration

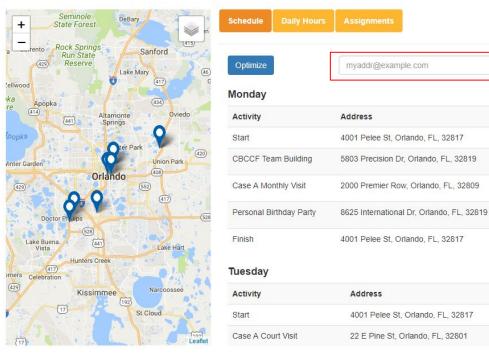
50 mins

60 mins

120 mins

Duration

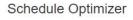
120 mins



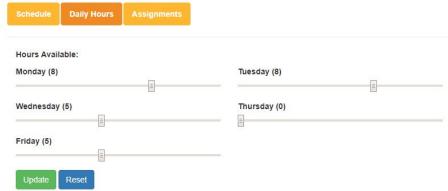
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Daily Hours

Map



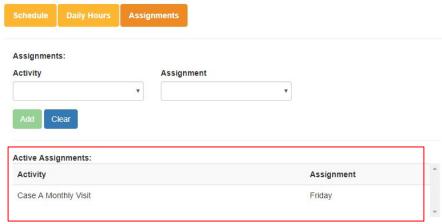




Assignments





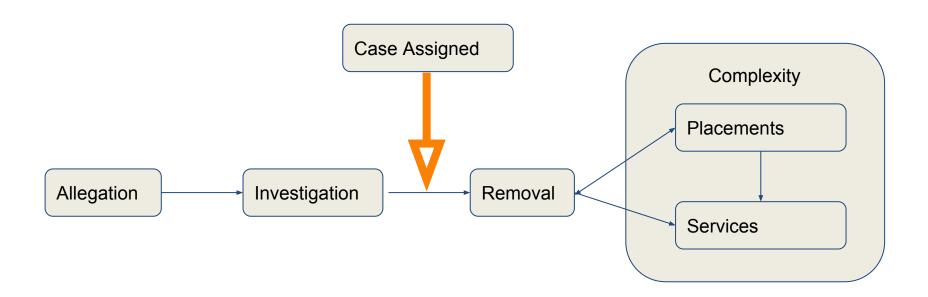


Case Complexity

Problem Statement

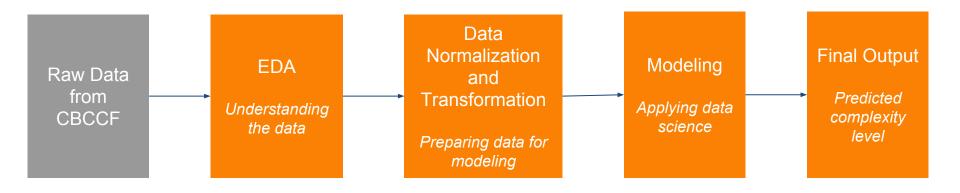
- Current assignment leads to imbalance
- Measuring case complexity will allow managers to balance case assignments
- Give specialized cases to specific people
- Group siblings together or geography

Timeline



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Our Process

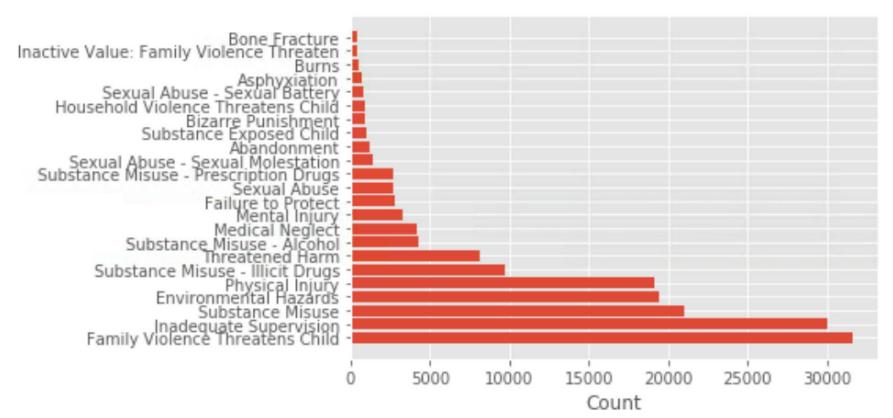




Exploratory Data Analysis

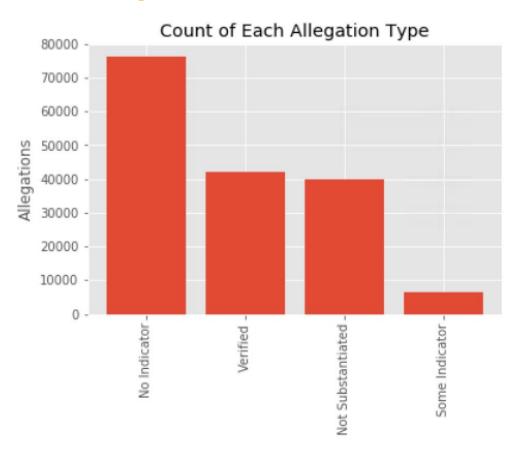
- Volume of Data
- Tables included:
 - Caseworker data
 - Case notes
 - Case notes person level
 - Allegations
 - Investigations
 - Placements
 - Reasons for Removal
 - Family connections

Allegations





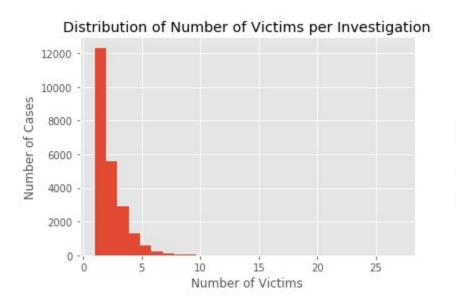
A quarter of allegations end up verified

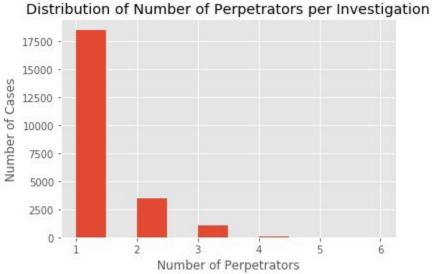




Most cases have 1 perpetrator and fewer than 5 victims

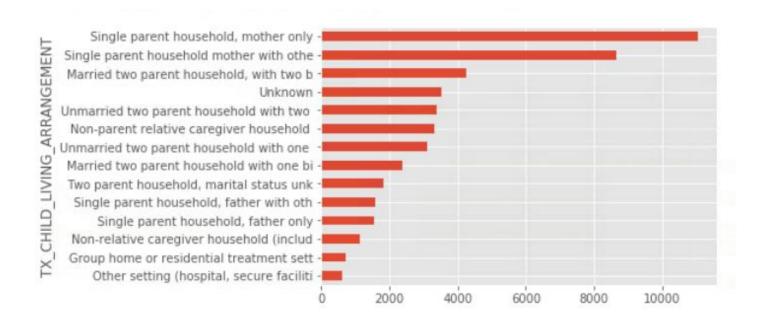
For Seminole County:





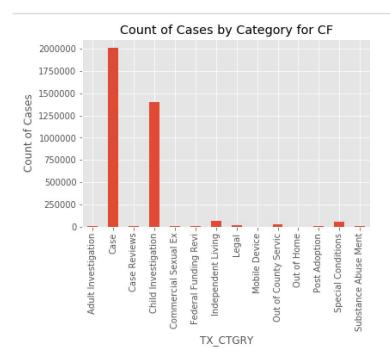


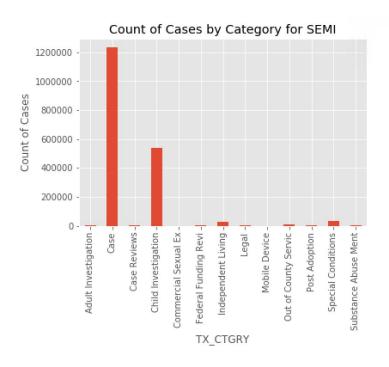
48% of investigations are made in to single parent homes



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Visualizing Categories

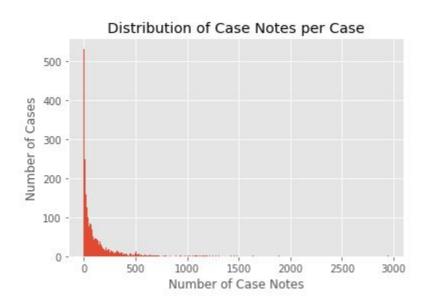






The median number of notes/case is 145

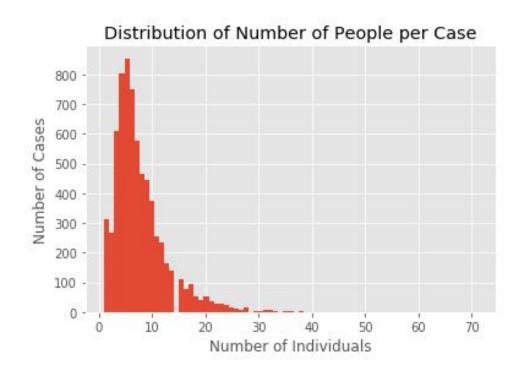
For Seminole County:





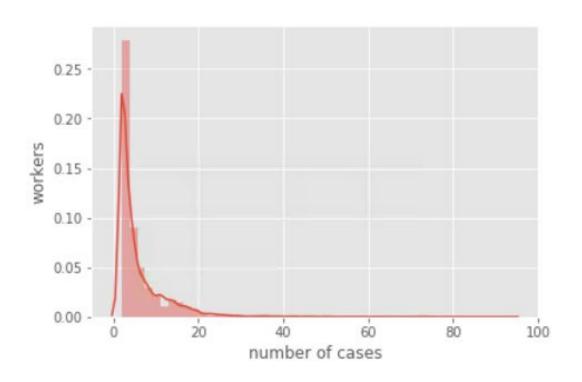
The median number of people/case is 10

For Seminole County:



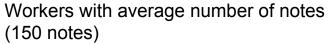


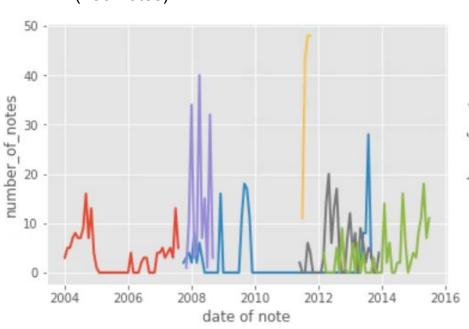
In a month workers log activity for 3 cases



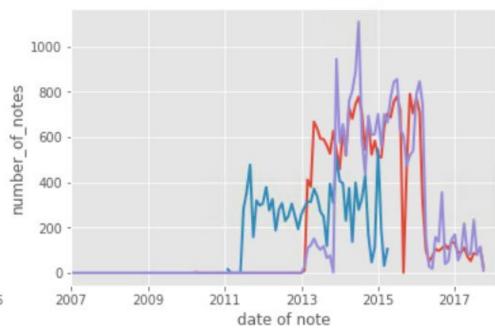


Workers log notes in bursts of activity (CF)





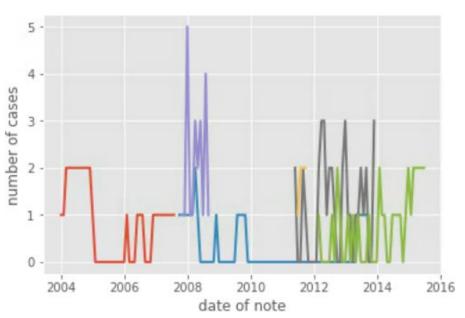
Workers with more than 10000 notes



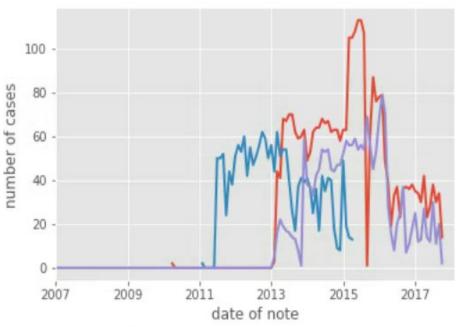


There is a huge range in case load (CF)

Workers with average number of notes (150 notes)

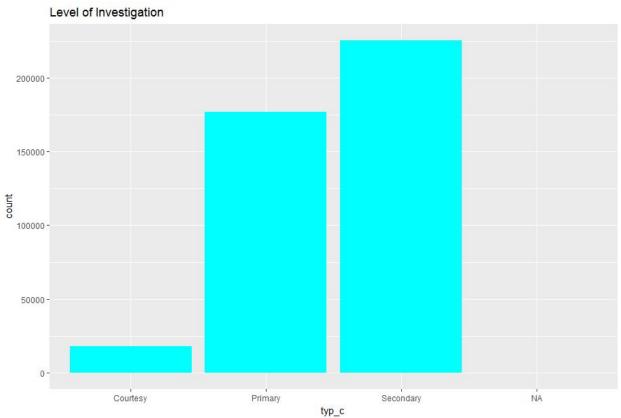


Workers with more than 10000 notes



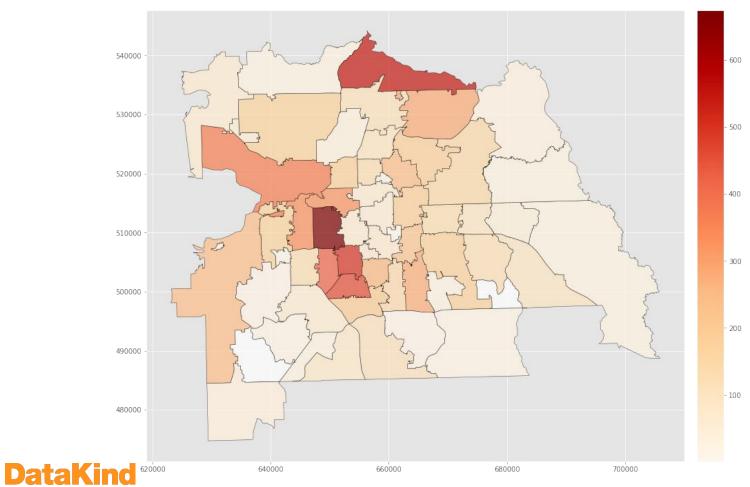


Level of Investigation

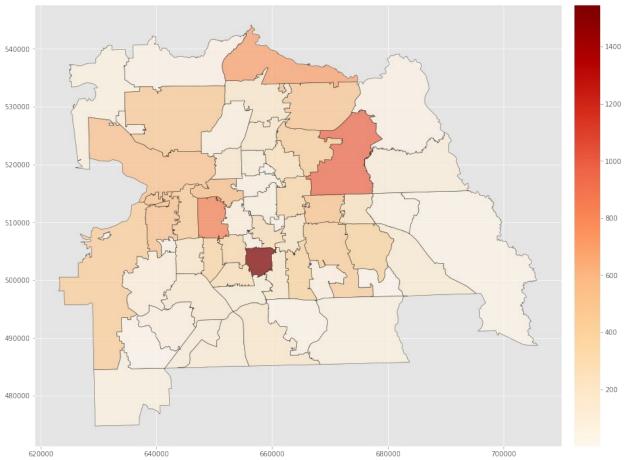




Removals by Zip Code

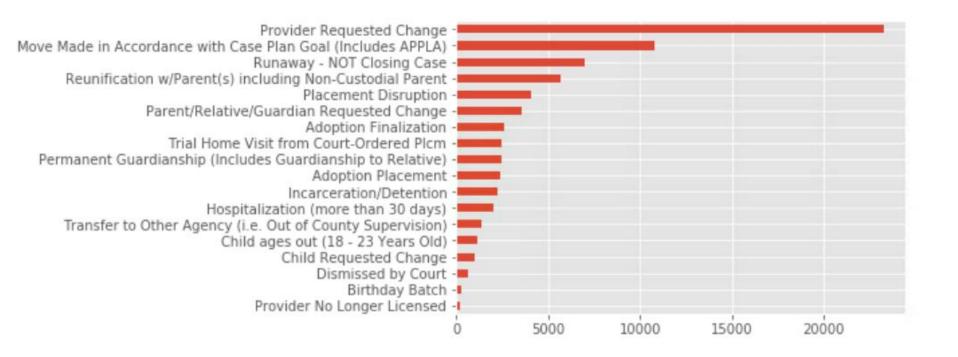


Placements by Zip Code



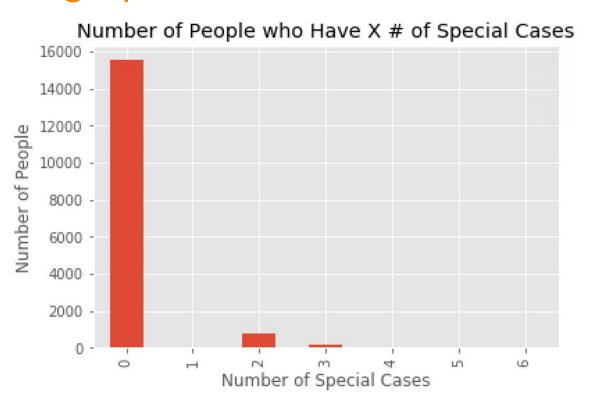


Common reasons for why placements end





Visualizing Special Cases





Questions at the time (2 months in)

- How can we find when a case was opened (and closed)?
- What does the end of a placement mean?
- How often do new cases come up?
- What time window should we predict complexity in order for it to be relevant during case assignments?
- With regards to case notes: 1) In a month workers log activity for ~3
 cases. Does this seem low? 2) Why are we observing spikes in log activity
 rather a uniform behavior? Are notes loaded in bulk?

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Current Uses of Machine Learning

- Danger Level at Allegation/Investigation
- Avoid High Risk Predictions

Tracing a Case

• Needed to visualize a case to get clear understanding

Fast Facts

Jan 6, 2009 - June 17, 2014

8 Individuals:

- 4 daughters
- 4 adults (?)
 - 2 parents, 1 maternal grandfather
 - 1 unidentified individual

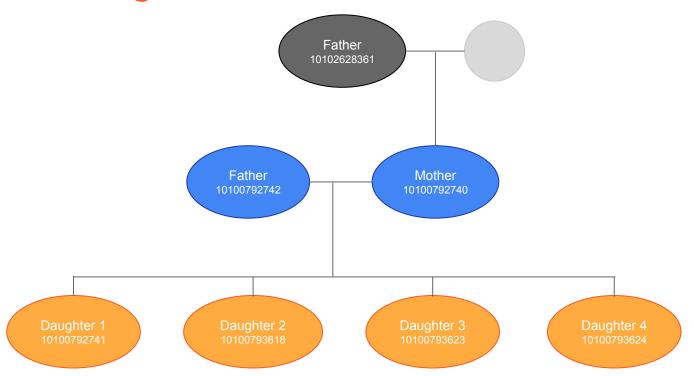
Allegations:

- Substance misuse
- Threatened harm
- Family violence threatens child

Reasons for Service:

- Inadequate supervision
- Incarcerated parent
- Alcohol abuse by the parent
- Drug abuse by the parent

Family Tree

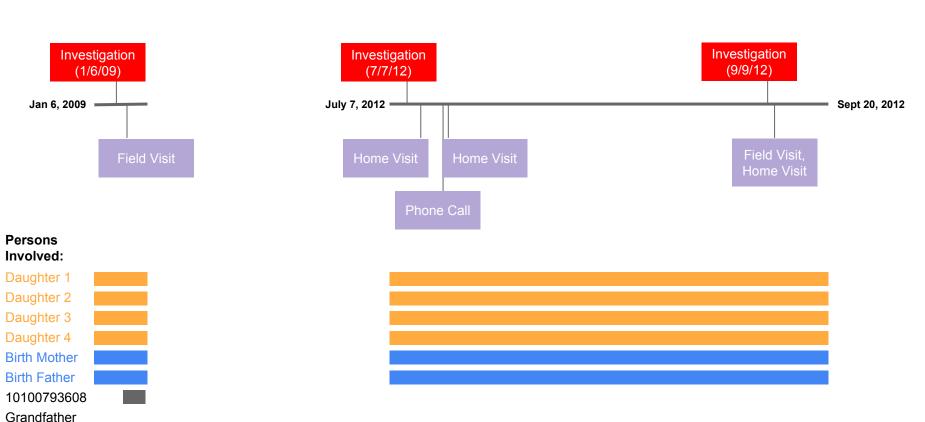


Fast Facts

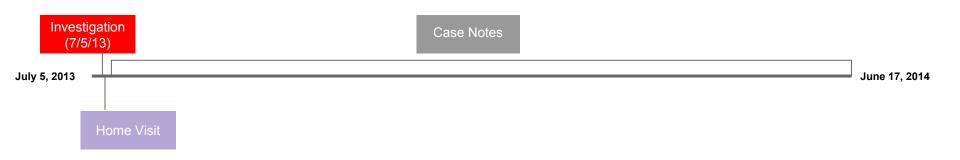
14 Caseworkers worked on the case from 7/20/2012 onwards

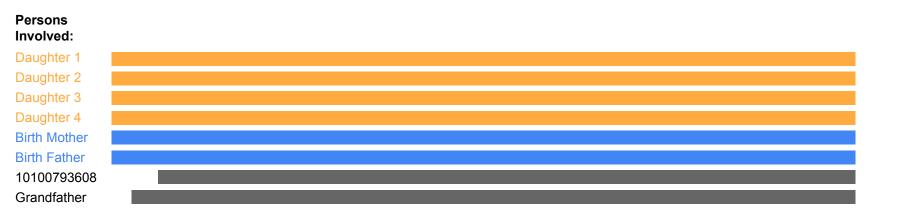
- 4 primary
 - o 9922807837: 7/20/12, 11/7/12, 11/9/12
 - o 9922808566: 7/20/12 11/9/12
 - 0 9922818164: 7/17/13 7/23/13
 - 0 9922825902: 7/23/13 6/30/14
- 10 secondary
 - 9922807837: 7/20/12
 - 9922824272: 7/16/13 9/3/13
 - 9922824273: 7/16/13 9/3/13
 - 0 1180040000: 7/16/13 6/30/14
 - o 9922767926: 7/16/13 6/30/14
 - 0 9922766481: 7/16/13 6/30/14
 - 9922790794: 7/19/13 6/30/14
 - 0 9922797904: 10/3/13 11/5/13
 - 9922823537: 11/18/13 11/27/13
 - 9922711699: 11/26/13 6/30/14
- 1 Out of County
 - 9922789900: 11/8/13 11/11/13

Timeline of the Case



Timeline of the Case (cont'd)





1/6/09 – Investigation # 1

Living Arrangement: married two parent household, with two b

CPS Type: In-Home

Response Priority: 24 hours

Allegations and investigated outcome:

Child	Allegation	Perpetrator	Outcome
Daughter 1	Substance Misuse	0	Some indicator
	Threatened Harm	0	No indicator
Daughter 2	Substance Misuse	0	Some Indicator
Daughter 3	Substance Misuse	0	Some Indicator
Daughter 4	Substance Misuse	0	Some Indicator

1/6/2009 - 1/10/2009

1 Investigation

1 Field Visit

7/7/12 – Investigation # 2

Living Arrangement: married two parent household, with two b

CPS Type: In-Home

Response Priority: 24 hours

Allegations and investigated outcome:

Child	Allegation	Perpetrator	Outcome
Daughter 2	Family Violence Threatens Child, Substance Misuse	0	Not Substantiated
Daughter 3	Family Violence Threatens Child, Substance Misuse	0	Not Substantiated
Daughter 4	Family Violence Threatens Child, Substance Misuse	0	Not Substantiated

7/7/2012 - 9/13/2012

- 1 Investigation
- 1 Home Visit to the Child's Current Residence
- 1 Home Visit to the Parent/Caregiver
- 1 Telephone Contact
- 1 Note to File

9/19/12 – Investigation # 3

Living Arrangement: Single parent household, father only

CPS Type: In-Home

Response Priority: 24 hours

Allegations and investigated outcome:

Child	Allegation	Perpetrator	Outcome
Daughter 2	Substance Misuse	Mother	Verified
Daughter 3	Substance Misuse	Mother	Verified
Daughter 4	Substance Misuse	Mother	Verified

9/19/2012 - 9/20/2012

- 1 Investigation
- 2 Field Visits
- 1 Home Visit to the Child's Current Residence

7/5/13 – Investigation # 4

Living Arrangement: Single parent household, father only

CPS Type: In-Home

Response Priority: 24 hours

Allegations and investigated outcome:

Child	Allegation	Perpetrator	Outcome
Daughter 2	Substance Misuse	Mother	Verified
Daughter 3	Substance Misuse	Mother	Verified
Daughter 4	Substance Misuse	Mother	Verified

7/5/2013 - 6/17/2014

- 1 Investigation
- 1 Home Visit to the Parent/Caregiver
- 46 Case Notes missing tx_type

Minimum Viable Product

- Simplest iteration possible
- See feasibility of machine learning (is this possible?)

Recap

- How do we measure complexity?
- Is it possible to give specialized cases to certain people?
- Is it possible to take geospatial information into account?

How to measure complexity

• Initial Idea: Predicting End Date

Problem Setup

How do we define "complexity"?

Our approach is to use activity as a proxy for a case's complexity.

- The activity rate is defined as the number of case notes entered in a 30 day period.
- We then convert the activity rate into a duration in hours using the estimated duration data provided by CBCCF

Why use an activity rate?

The advantages of using a activity duration:

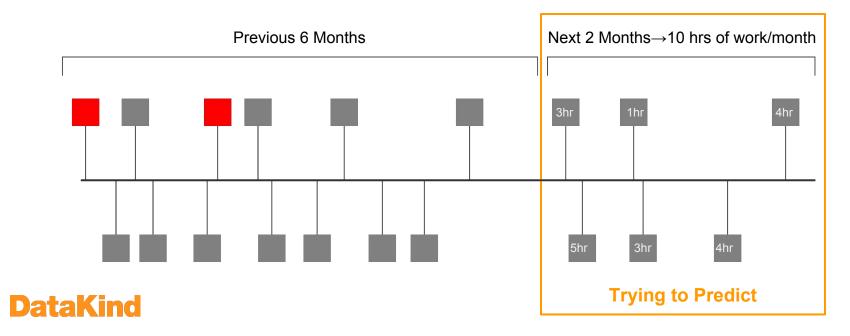
- Easy to understand
- Actionable
- Less influenced by a caseworker's ability
- Somewhat independent of the number of caseworkers assigned to a case, since we can't fully segment out work done by individual case workers on a case

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Problem Setup

How are we calculating complexity? (What we are predicting...)

For each case, we will be using data attributes from the previous 6 months of the case to predict the average level of activity for the next 2 months.

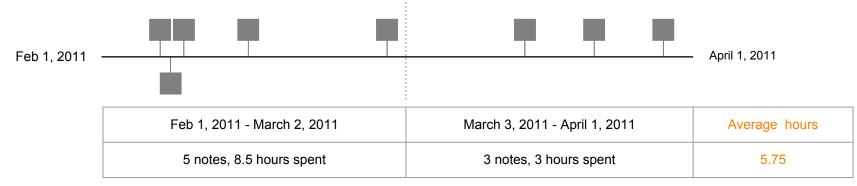


Model predicts: average hours/month

Counting Average Activity

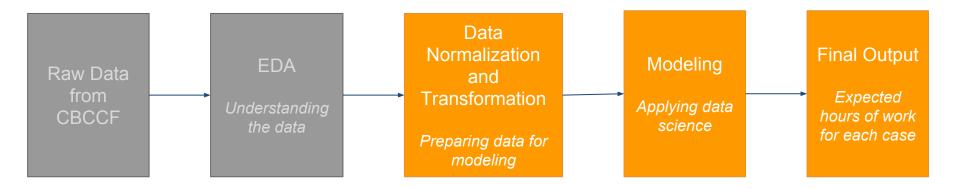
- Calculated the average number of hours spent in the seventh and eighth months of each case
 - This is what we will be trying to predict

An example: Case 10001186814





Our Process



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Representative case dataset

Determining Eligible Cases:

- At least 3 months long but no longer than 18 years
- Minimum of 5 case notes
- Caseworkers with 10 or more handled cases
- Total of 8,907 cases from CBC jurisdiction included



Activity in the representative case dataset

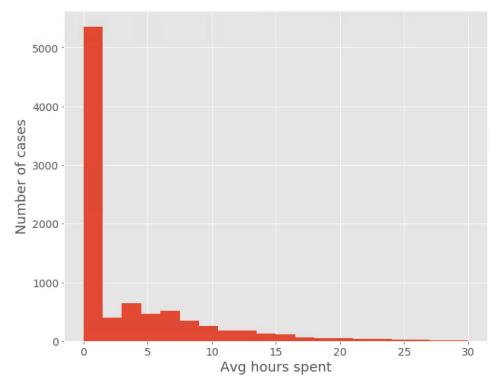
Mean: 3.56 hours

Median: 0.0 hours

(5258 cases)

Max: 159. Hours

(Case id: 10001763991)





Data we normalized and aggregated

 Investigations: Living arrangement Access type CPS type Determination Response priority 	 Placements: Filter to placement -related services Number of placements Number of placement zip codes Placement duration 	 Removals: Manner of removal Family structure Discharge reason Reason for service
Reason for Service: • Reasons for why child is considered at-risk	Notes:Note categoryType of work being done	

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Count vectorizing: what the models sees

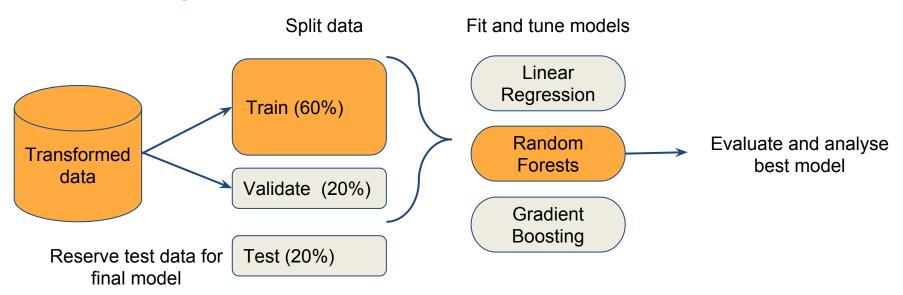
Activity Type	Date
Investigation	7/7/12
Home Visit	7/8/12
Telephone Contact	7/16/12
Home Visit	7/17/12
Home visit	9/13/12
Investigation	9/19/12
Placement	9/20/12
Placement	9/25/12

Count each unique activity

Investigation	Home Visit	Telephone Contact	Placement
2	3	1	2



Modeling Process

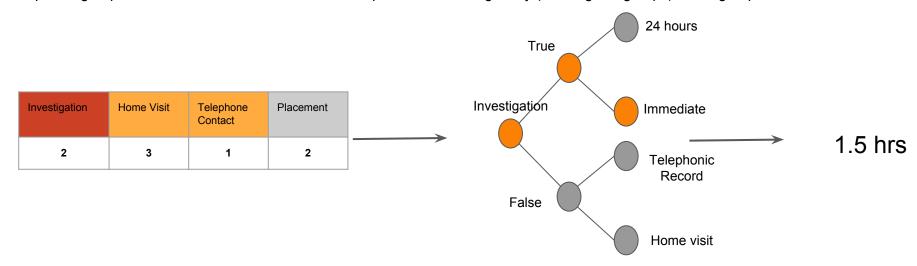


Mean absolute error (MAE): actual hours spent - predicted hours



Decision tree

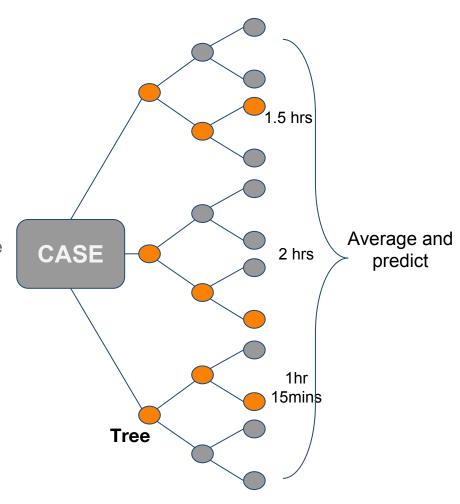
Decision Tree Classifiers: Decision Tree Classifiers are used to predict outcome of an observation. The model maps information about a specific observation to a specific outcome. Such models are popular because of their ease of interpretation and simplicity of application. The model can be thought of as a series of if/then statements that allow for prediction of an observation belonging to a specific group/outcome. The if then statements attempt increase homogeneity (creating like groups) within groups.



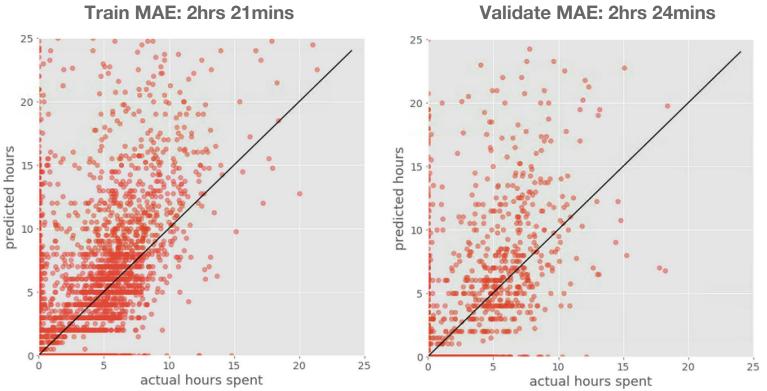


Random forests

- Random Forest Classification uses the Classification Tree approach but instead of estimating a single tree, multiple trees are created (i.e. tree -> forest).
- This algorithm is referred to as an "ensemble" method as it brings together multiple models to predict a single outcome. Each observation traverses each tree and each tree "casts a vote" the proportion of votes are used to create prediction.
- This algorithm has strong predictive power, can handle missing or incomplete data and is relatively easy to implement.
- One downside is that the model logic is difficult to interpret, but one can generate a list of the most important features in the model.

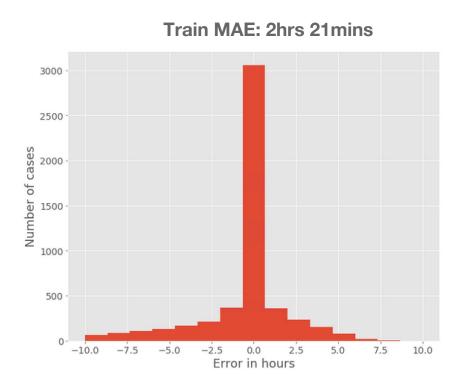


On average we have an error of 2hrs 24mins

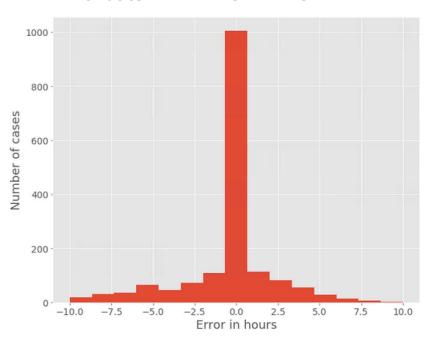




Frequency distribution of errors



Validate MAE: 2hrs 24mins





20 most important features

Investigations:

 Child Investigation Initial Face-to-Face

Placements:

- Count of placements
- No of_distinct_zips
- total_days_per_placement
- placement_cat_FosterHome
- placement_cat_Relative

Removals:

Reason for Service:

- Closing-Services
- Child Intake
- In-Home

Notes:

- Case Note to File Legal
- Case Field Visit Other
- Case Note to File General
- Case Telephone Contact
- Case Court Judicial
- Case Reviews Supervisory
- Case Home Visit Child's Current Residence

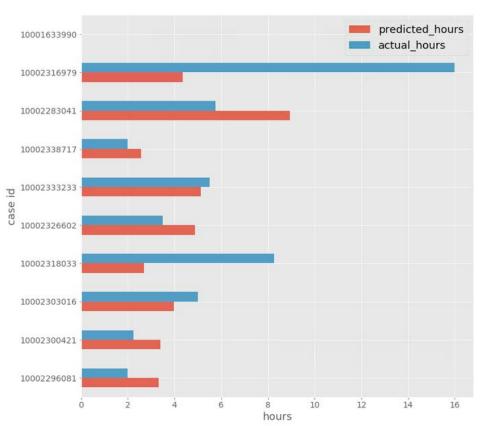
Notes:

- Case Service Provider Contact
- Case Office Visit
- Case Other
- Case Home Visit -Parent/CG Removed From

Example use case

- Caseworker with 9922849324 was assigned to 10 cases as a primary starting between January 2017 and April 2017
- On the 1st, Aug 2017 a Supervisor wanted to estimate how many hours would this case worker likely work in the next two months.
- The model was used to predict the hours they would spend on 1st, Aug 2017 for the next two months.
- At this point in time every case had been active for at least 3 months
- We then analyzed what features the model saw for each case

What did we predict?





Case 3233, Predicted: 5.1hrs, Error: 0.4 hrs

Investigations:

- Single parent household, mother only
- 24 Hours
- Child Investigation Commencement (Initial)

Placements:

- placement_cat_Relative,
- placement_flg_Y,
- Total_days_per_placement
- no_of_distinct_zips

Removals:

Reason for Service:

- Closing-Services
- Child Intake
- In-Home
- Closing-Open Ongoing Case Management Ser,

Notes:

- Case Telephone Contact
- Case Note to File Legal
- Case Note to File General
- Child Investigation
- Telephone Contact
- Child Investigation Note
- Child Investigation Home Visit - Parent/Caregiver,

Notes:

- Case Court Judicial,
- Case Visitation -Parent/CG Removed
- Case Service Provider Contact
- Case Home Visit -Child's Current
 Residence

Case 8033, Predicted: 8.25hrs, Error: 5.5 hrs

Investigations:

- Single parent household, mother only
- Immediate,

Placements:

- placement_cat_Medical -Hospitalization
- placement_cat_Relative
- placement_flg_Y
- Total_days_per_placements
- no_of_distinct_zips

Removals:

Reason for Service:

- Child Intake, In-Home,
- Closing-Open Ongoing Case Management Ser

Notes:

- Child Investigation Home Visit - Child's Current Residence
- Child Investigation Field Visit
- Case E-Mail Contact
- Child Investigation
 Commencement (Initial)
- Case Note to File General

Notes:

- Child Investigation Telephone Contact
- Child Investigation Note to File - General
- Case Supervisory Consult
- Case Home Visit -Child's Current Residence

Case 6979, Predicted: 8.9hrs, Error: -3.2 hrs

Investigations:

- Single parent household mother with othe
- 24 Hours, Immediate

Placements:

Removals:

Reason for Service:

- Child Intake,
- Closing-No Services,
- Closing-Open Ongoing Case Management Ser
- fl_drg_abs_prnt

DataKind

Notes:

- Child Investigation Field Visit
- Case E-Mail Contact
- Case Office Visit
- Case Visitation Relative,
- Case Telephone Contact
- Case Note to File General,
 Child Investigation

Notes:

- Child Investigation Home Visit -Parent/Caregiver
- Case Visitation -Parent/CG Removed From
- Case Service Provider Contact, Case Home Visit - Child's Current Residence,

Post-MVP Model Improvements

Sampling changes

Include activity from the middle and end of cases, not just the beginning

Model tuning

Make adjustments to improve model performance

Model interpretability

Shed light on why the model made the predictions it did

Including activity from the middle and end of cases

Before:

All activity was drawn from the first six months of a case

Now:

1/3 of sample is from the first six months

1/3 of sample is from the last six months

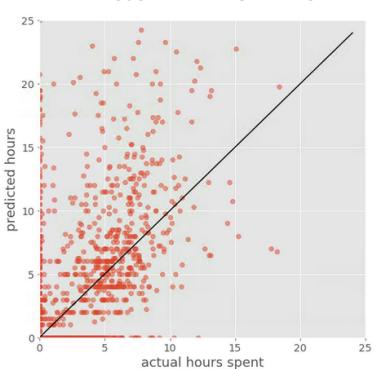
1/3 of sample is randomly drawn from the middle



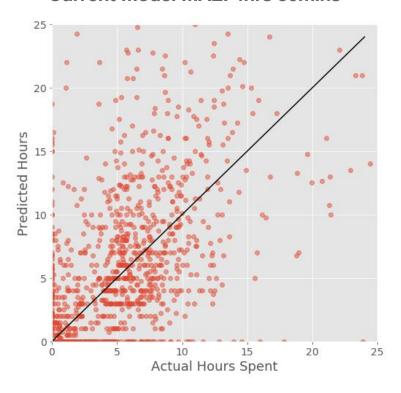


Improving Model Accuracy

MVP Model MAE: 2hrs 24mins

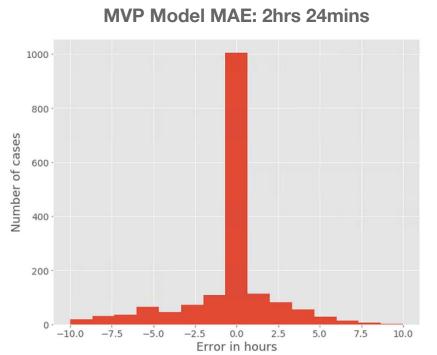


Current Model MAE: 4hrs 30mins

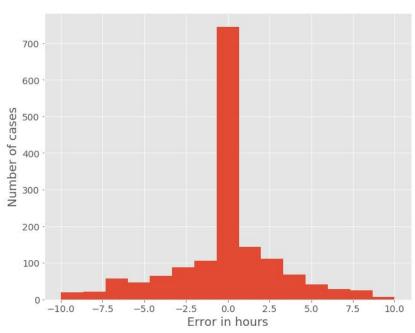




Frequency distribution of errors

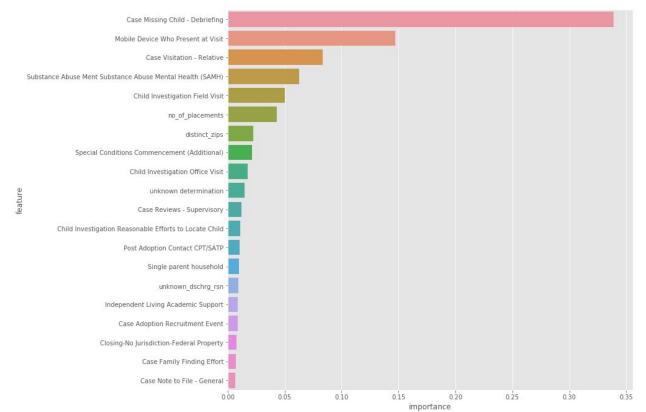


Current Model MAE: 4hrs 30mins





Adding model interpretability





Adding individual case-level interpretation

Case 5434

Predicted: 2.3 hours Actual: 2.0 hours

Most Significant Features:

- Mobile Device Who Present at Visit
- 2. Case Missing Child Debriefing
- 3. Number of zip codes
- 4. Case Visitation Relative
- 5. Number of placements
- 6. Substance Abuse Ment Substance Abuse Mental He
- 7. Child Investigation Field Visit
- 8. Child Investigation Staffing Multidisciplinary
- 9. Special Conditions Staffing Red Flag
- 10. Case Family Finding Effort

Adding individual case-level interpretation

Case 49104

Predicted: 2.6 hours Actual: 14.5 hours

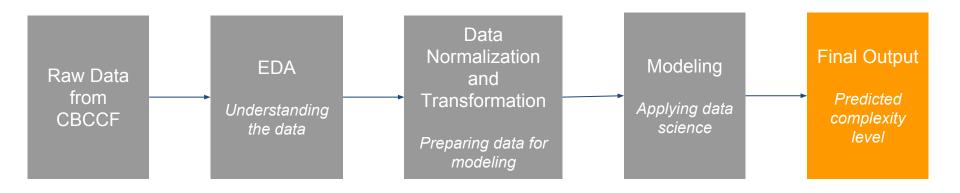
Most Significant Features:

- Case Visitation Relative
- 2. Case Missing Child Debriefing
- 3. Fl_alc_abus_prnt
- 4. Post Adoption Contact CPT/SATP
- 5. Mobile Device Who Present at Visit
- 6. Number of placements
- 7. Child Investigation Field Visit
- 8. Substance Abuse Ment Substance Abuse Mental
- 9. Number of zip codes
- 10. Closing-No Jurisdiction-Federal Property

Major Error

• Some of the cases were being labelled with 0 activity

Our Process



What We Are Predicting

What is the output of our model?

Our model offers a prediction of the level of complexity (low, medium or high) over the next 60 days for each case based on case attributes and activity from the previous 6 months.

- Low complexity = 0 4 hours per month
- Medium complexity = 5 16 hours per month
- High complexity = over 16 hours per month

Our model also attempts to shed some light on the factors that led to its predicted level of complexity.

Our Case Dataset (n = 9,280)

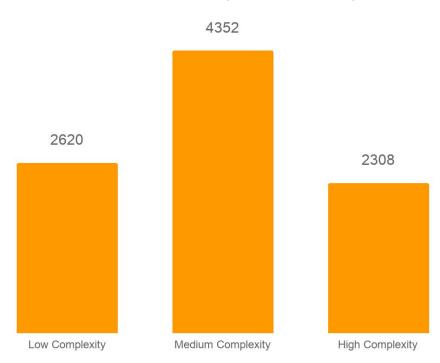
Eligible Cases:

- At least 3 months long but no longer than 18 years
- Case started after Jan 1, 2000
- Case has at least 5 notes.

Case Activity:

- Mean: 14.5 hours
- Median: 7.5 hours
- Maximum: 380 hours

Number of Cases by Complexity Level



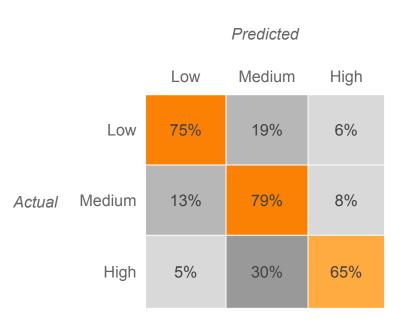


Our Model Features

 Investigations: Living arrangement CPS type (in home or not) Determination Response priority 	 Placements: Number of placements Number of placement zip codes Placement duration 	 Removals: Manner of removal Family structure Discharge reason Reason for service
Reason for Service: Reasons for why child is considered at-risk	Notes:Note categoryType of work being done	

Our Prediction Accuracy

Confusion Matrix:

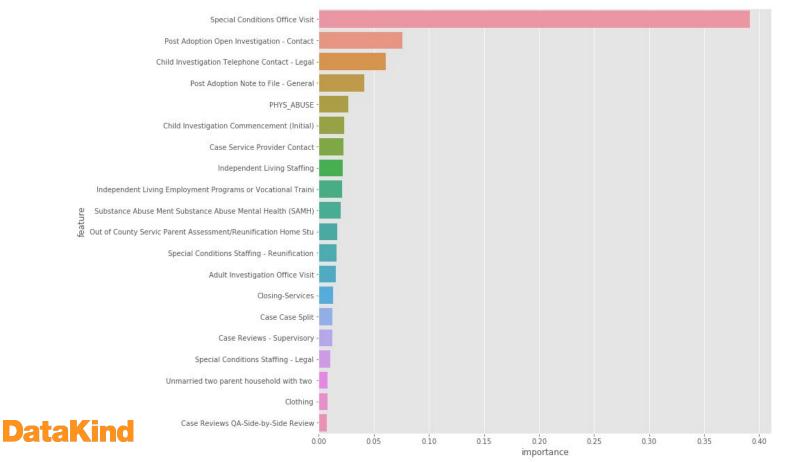


Overall, our model predicts the correct level of complexity 72% of the time (*precision*) and correctly classifies the complexity of 72% of all cases (*recall*).

	<u>Precision</u>	Recall
Low	0.69	0.71
Medium	0.71	0.79
High	0.78	0.59
Average	0.72	0.72



The Most Important Features



Example Use Case

- A supervisor wants to estimate the level of complexity for Case 9999797272 for the next 60 days
- Over the previous 6 months, Case 9999797272 had the following activities:
 - 16 Special Conditions Facsimile
 - 3 Independent Living Budget and Financial Management
 - 2 Adult Investigation Home Visit VA's Current Residence
 - 2 Child Investigation Office Visit
 - 2 Out of County Service Home Study
 - 2 Residential
 - 1 Special Conditions
 - 1 Case Home Study
 - And 9 other activities



Example Use Case

Model Prediction	Actual Hours Spent	Model Confidence
Medium Complexity	7	83%

The most important features identified for this case:

- Case Staffing Reunification
- Clothing
- Special Conditions Commencement (Initial)
- Bed Hold
- Special Conditions Commencement (Additional)
- Special Conditions Staffing Case Transfer/ESI
- Child Investigation Note to File Interstate Compact
- Special Conditions Home Visit Parent/Caregiver
- Case Visitation Parent/CG Removed From
- Case Visitation Sibling



Accomplished

Complexity Model

Did not address:

- Caseworker specialty
- Geography (kind of)

Lessons Learned

- Modular Code (Faster Iterations)
- Partner with Strong IT background
 - Talk about deployment early
- Communicating Predictions
- Maybe should've considered the relationship of two projects more
- Experience with other stacks

Getting Involved in DataKind

- Apply Online
- Opening of Projects Varies

Volunteer Opportunity

The Problem

When a child is removed from their families, the goal is to place them in the best home first and for the child to remain in that same home for the duration of their time in foster care. Kids that move from foster home to foster home multiple times end up spending more time in foster care, have academic (many times school moves also occur) and attachment issues. See the Casey Family Foundation brief on Placement Stability for more information (https://www.casey.org/placement-stability-impacts/)

Currently, there are mechanisms in place (Placement Stability Staffings) to help a foster home when the placement is breaking down, but in many cases, by the time the foster parent requests a Placement Stability Staffing, it is too late.

Volunteer Opportunity

The Challenge

Using sentiment analysis on case notes (and possibly other data points), can we predict a placement disruption before the foster parent requests a change. If so, we can suggest a Placement Stability Staffing or other supports before the foster parent needs to ask.

We hope that this will improve stability and ultimately lead to better outcomes for the kids we care for.