

Combining Clustering and FSM models for Adaptive Level Generation

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Abstract—

I. INTRODUCTION

problem: levels are finite
space constraints and design is costly
to solve it: algorithms
that is, pcg
a way to generate levels
rules
is hs been used for all these things in these games
like minecraft
or diablo
even tetris!
adaptive: why?
because it tailors better to tastes
advantages? personalization empowers bonding
hey i could probably find some sweet cites for this
it has been used here and there
sassy, brain training
or, hey, l4d2
our case: the mario competition
what have people used?
passes -> but no link
clusters -> that's cool, but bad random
hopper -> yeah that's cool
so we'll try to do both and cool
have one process guide the other
and an open architecture ripe for flexibility
we'll do this by separating functionality in layers
and encoding levels in transformable ways
urgh this is bad
retake it once the rest is done
you'll have a clearer idea

II. BACKGROUND

mario is a game about platforms and jumps
and enemies and coins
and you have to get to the end
the mario ai championship is about
plenty of stuff
but the level generation track is about
forming the best level there was
the framework is based on infinite mario by notch
which is

top notch
haha
so yeah
our goal is to make a good architecture for it
that can be flexible, random every time, adaptive, and not
too tailored to the game
how did we do it?
keep reading

III. METHODOLOGY

this is our architecture
there are FOUR LIGHTS
but three layers
two mostly
layer figure
top takes care of identifying the user
how does he like it
second, generating the level
the user identification guides the process
because
that's the way
u-hu, u-hu
he likes it
u-hu
for the top layer we decided to cluster
it gives a point of reference
he prefers this way, in comparison to others
judging in a vacuum is hard
also makes them feel special
the second, grammars
levels are a succession of repeated basic symbols
also, what comes now relies on what came before
we can model that as a directed graph
an automata, for example
built from the grammar
we can make new "paths", signifying possibilities, parts
of the screen
by adding new states or links
automata are flexible
and expressing them as grammars, quite intuitive
they rock
okay then

IV. CLUSTERING

we want to base player profiling based on data
we think it's more faithful
player's don't know what they want
and badly estimate themselves

cite here
 or at least steve jobs quote
 so let's take data
 mario stores two kinds of data
 metrics and logs
 we'll focus on the metrics
 why? because we don't want specific timestamped events
 but overarching behavior
 so mario challenge stores all this data on the player
 they refer to this categories: deaths, blah
 we are going to try to infer the different existing styles
 so we'll take one sample per person, for these reasons
 we need to provide them with comparable challenges
 for these reasons
 once gathered, we will try to find groups of palayers with
 similar behaviour
 we want three
 why three? they are everywhere
 experience proves it, but also
 the speeder prototype comes from speedruns, light-and-
 nimble characters, aggressive strategies
 explorers come from detail-intensive games, hevy ar-
 moured charactered, controllish, slow plans
 they form a spectrum, with two extremes
 so we decided to model with three, allocating for a mixed
 intermediate position
 also, not a lot of samples
 now to choose the algorithm
 we decided to rely on the weka library
 its cool and awesome and proven. source:
 weka offers these algorithms that are the most commonly
 used.
 we'll use this metric to compare them
 since it's density blah
 we need to blah
 we could add a parameter by the user's preferred difficulty
 but that's intrusive
 and they are bad at it anyway
 if you need to choose it before palying, you have no idea
 if you've alrady played, we have the metrics thankyou-
 verymuch
 ok so we have the clusters
 stored safely somewhere
 this process is made so it can be run offline
 the clustering of the data, i mean
 so then a player comes
 and we assign them one of those clusters

V. GRAMMARS

whew that was long
 ok so now to fsms
 when we assign the player to a cluster, we implicitly assign
 him a grammar
 since each cluster has a grammar associated
 the grammars are called schematics and describe how to
 build the level
 in a pseudo-bnf way

take a look at this simple example
 HERE BE FIGURE
 the derivation rules have weights
 a heavier rule has more probabilities of being chosen
 the schematics are context-free grammars
 and expands on the right
 so in the example we start the level
 and we can have a pipe or a coin or flat
 and we loop
 (ok explain this a bit longer)
 the schematic is designed to be infinite
 to be able to generate levels of any length
 this is a sample of a trace
 BLA PIPE COIN RICK ROLL
 the schematics are done by hand
 the terminals are predefined
 they are the available "chunks" that the system can build
 for example, a small gap
 or a pipe
 or a couple of blocks
 in this case we made every chunk of the same length, two
 we wanted the smallest possible to make it more flexible
 the arching structure can be obtained with the rules of the
 schematic
 forcing pieces to be together
 but if we wanted to control little details, we need little
 chunks
 but we couldn't do one, because pipes for example are two
 blocks
 and gaps of length one are not very mario-like
 we needed chunks of the same size so we had a calculable
 relation
 between chunks used and level length
 for the number of iterations
 we want this length, ok so we iterate this n of times
 also, levels of same length, same number of states
 so we can compare them
 the choice of cluster from the step before
 determines what schematic we read
 each one has derivations and weights tailored to their needs
 but, as we said, same nonterminals
 alright so we read the schematic to construct an automaton
 representing these transitions
 we use the parse2 library
 we defined a grammar for the schematics
 then parse2 reads the schematics and transform the rules
 into transitions
 with their weights of course
 the resulting graph
 can be seen also as a tree with cycles
 see the figure for the tree for our example from before
 HERE
 in this tree model, we'd be traversing it depth-first
 we store the terminals we follow, forming a string of
 chunks to place
 we call this string a "genotype", because it represents the
 level

two equal genotypes represent the same level
an executor later traverses the chunks
from starting node
chooses a random transition
pushes the derivation chain into the stack
then pop top node
if it's terminal, we add it to genotype
if it's nonterminal, we pick a random transition and push
it to stack
always proceed with the leftmost element of the chain first
because we build the level from left to right

VI. RESULTS

god damn that was a lot to tell
so what happened
we used social networks, like twitter, facebook and reddit,
and word of mouth
to get the data
117 entries
each person was given a package with ready-to-play game
the game would provide a regular random level from the
framework
they were asked to just play, with no other worries about
completing the stage, or training before hand
and send back the results
we fed into weka
but we filtered some parameters
because the default stage
has no gaps, or enemies other than goombas
so some parameters were always zero
those parameters:
we used those, though:
then we tried the five clustering algorithms with default
settings
except three clusters as explained
(why only the five? er...)
you can see the results of the log likelihood here
em rocked
the three clusters found represented the profiles we hoped
check this sweet table
isn't it gorgeous
plus they are well separated
check this graph
you like it huh?
speeders like this
explorers like that
and my milkshake brings all the boys to the yard
so we made some grammars
here, take a ride in my intermediate
here's the diagram
and a sample trace for good measure
ok a screenshot too
let me explain some design choices
like, this branch here

VII. CONCLUSIONS

the separation of layers means that the way you identify
users
is not thoroughly linked to how you use that information
to guide the building
grammars meant
that we used the same basic chunks
in different ways each
tailoring for styles
grammars are easily intuitive and modifiable
you can control what goes next to what
and how often
and offer choices, different "paths" for the build
we also produced a way of identifying and storing a level
by its genotype of chunks
(did you define chunk? go and do it!)

Future work

we hope
to evaluate this system
with some group control
also
genotypes allow us to use genetic techniques
for example
if the clustering layers defines a player
instead of one membership
with shared membership
we can generate levels from different automata
and mix them
we can also alternate picking derivation rules between
grammars by those same percentages

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