Gambler's Ruin Problem

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Gambler's ruin problem 1

In Gamblerruin.thy, we will construct the formalization of a specific random walk model coordinated with gambler's ruin problem.

1.1 Theory Infinite_Coin_Toss_Space

In order to construct the formal method in gambler's ruin problem, we start with the existing formalization in the Theory Infinite_Coin_Toss_Space which constructed the probability space on infinite sequences of independent coin, tosses.

The only concept need to be elaborated is bernoulli. 'a stream is a type of infinite sequence with all element of type 'a. The bernoulli stream is a stream measure which has the space composed of all the elements of type bool stream, measurable sets filled with all the subset of its space. under this specific measure, all the possibility of occurrence of elements in a specific set A can be described as the measure value of A if and only if A is in the measurable sets of this bernoulli stream. In fact, it was set up by producting countable measure of boolean space with measuring $\{\text{True}\}$ to p and $\{\text{False}\}$ to 1-p.

1.2 Gambler model

```
fun (in infinite-coin-toss-space) gambler-rand-walk-pre:: int \Rightarrow int \Rightarrow int \Rightarrow (nat \Rightarrow bool stream \Rightarrow int) where base: gambler-rand-walk-pre u d v 0 w = v| step1: gambler-rand-walk-pre u d v (Suc n) w = ((\lambda True \Rightarrow u | False \Rightarrow d) (snth w n)) + gambler-rand-walk-pre u d v n w
```

```
fun (in infinite-coin-toss-space) gambler-rand-walk:: int \Rightarrow int \Rightarrow int \Rightarrow (enat \Rightarrow bool stream \Rightarrow int) where gambler-rand-walk u d v n w = (case n of enat n \Rightarrow (gambler-rand-walk-pre u d v n w)|\infty \Rightarrow -1)
```

The function $gambler_ran_walk$ extends the fourth parameter by adding ∞ as new input. The reason why we define it is that we found it very tough to describe the position where the specific random walk stops, for the first time, by reaching the threshold if natural number is the only allowed input as what $gambler_ran_walk_pre$ defines. Since some infinite random walks will never stop, we must allocate ∞ as the output coordinated with that non-stop case and extend the type of steps from nat to enat. But if someone wants to base their further analysis on our endeavor here, please be cautious of or even avoid discussing the case that initial number and target number is negative since we map ∞ to -1. The lemma exist demonstrates that non-stop random walk will never succeed in reaching the target, which is the best explanation why we allocates -1 as the output of ∞ in $gambler_ran_walk$.

```
locale gambler-model = infinite-coin-toss-space +
fixes geom-proc::int \Rightarrow bool stream \Rightarrow enat \Rightarrow int
assumes geometric-process:geom-proc init \ x \ step = gambler-rand-walk \ 1 \ (-1) \ init \ step \ x
```

begin

1.3 Basic functions

Here we define the all basic functions which will play an invisible role in the further probability analysis. you can just focus on the lemmas and functions where we comment

```
definition reach-steps::int \Rightarrow bool stream \Rightarrow int \Rightarrow nat setwhere reach-steps init x target = {step::nat. geom-proc init x step \in {0,target}}
```

reach_steps describes all the steps where the input random walk reaches the threshold 0, target

```
fun infm::nat\ set \Rightarrow enat\ \mathbf{where} infm\ A = (if\ A = \{\}\ then\ \infty\ else\ \square\ A) infm A = \text{iff}\ A = lemma only\text{-}inf\text{-}infm: assumes A \neq \{\} shows infm\ A \neq \infty \langle proof\ \rangle
```

```
fun stop-at::int \Rightarrow bool stream \Rightarrow int \Rightarrow enat where stop-at init x target = (infm (reach-steps init x target))
```

 $stop_at$ describes the first step in the $reach_steps$ sets, which means exactly the stopping point in gambler's ruin problem. Be careful, Here the type of output has been extended to enat, which means stopping point will be ∞ (equivalent to non-existence)

```
fun success::int \Rightarrow bool stream \Rightarrow int \Rightarrow boolwhere
success init x target = (geom-proc init x (stop-at init x target) = target)
```

success describes the random walk reaching the target number rather than ruining at stopping point

1.4 Important intermediate conclusions

1.4.1 Successful random walks never stop at ∞

Once we set target to be positive, the weird situation where random walk succeeds at ∞ will disappear

```
lemma exist:

fixes init::int and x and target::int

assumes 0 \le init \ init \le targetsuccess \ init \ x \ target

shows stop\text{-}at \ init \ x \ target \ne \infty

\langle proof \rangle
```

1.4.2 The way we count never change the amount got through specific random walk

```
lemma pre1: \bigwedge x \ n. \ snth \ x \ (n+1) = snth \ (stl \ x) \ n \ \langle proof \rangle
```

lemma additional states that the reaching number doesn't change if we want to calculate from the second step

```
lemma additional1:let init' = geom-proc init x 1 in geom-proc init' (stl x) n = \text{geom-proc init } x \text{ (Suc } n) \langle \text{proof} \rangle
```

1.4.3 The way we count never change whether the random walk succeeds

```
\begin{array}{l} \textbf{lemma} \ set\text{-}up\text{-}Inf: \\ \textbf{fixes} \ A \ \textbf{and} \ a::nat \\ \textbf{assumes} \ \bigwedge b::nat. \ b \in A \Longrightarrow a \leq b \ a \in A \\ \textbf{shows} \ a = Inf \ A \\ \langle proof \rangle \\ \\ \textbf{lemma} \ Inf\text{-}property: \\ \textbf{fixes} \ a \ \textbf{and} \ A \\ \textbf{assumes} \ a = Inf \ A \\ \textbf{shows} \ \bigwedge b::nat. \ b \in A \Longrightarrow a \leq b \\ \langle proof \rangle \end{array}
```

 $conditional2_pre$ states that stopping point doesn't change if we calculate from second step

```
lemma conditional2-pre:
fixes init' and Ar and Al
assumes init' = geom\text{-}proc\ init\ x\ 1
Ar = reach\text{-}steps\ init\ x\ target
Al = reach\text{-}steps\ init'\ (stl\ x)\ target
0 < init
init < target
shows stop\text{-}at\ init'\ (stl\ x)\ target + 1 = stop\text{-}at\ init\ x\ target
\langle proof \rangle
```

conditional2 states that whether a random walk succeeds or not doesn't change if we calculate from second step

```
lemma conditional2:

fixes init x target

assumes init' = geom-proc init x 1

0 < init

init < target

shows success init' (stl x) target \longleftrightarrow success init x target

\langle proof \rangle
```

1.4.4 The change of initial number

if first step is true, then we add 1 to initial number

lemma fst-true-plus-one:
fixes init x target
assumes init' = geom-proc init x 1shd x = True
shows init' = init + 1
⟨proof⟩

if first step is False, then we reduce 1 to initial number
lemma fst-true-plus-one-false:
fixes init x target
assumes init' = geom-proc init x 1shd x = False
shows init' = init − 1
⟨proof⟩

1.4.5 The way we count never change the successful random walk set

the set where all random walks in it succeeds and their first step are True doesn't change if we calculate from second step

 $\mathbf{lemma}\ conditional\text{-}set\text{-}equation:$

```
fixes init target assumes 0 < init \\ init < target shows \{x::bool\ stream.\ success\ init\ x\ target \land shd\ x = True\} = \\ \{x::bool\ stream.\ success\ (init+1)\ (stl\ x)\ target \land shd\ x = True\} \\ \langle proof \rangle
```

the set where all random walks in it succeeds and their first step are False doesn't change if we calculate from second step

```
lemma conditional-set-equation-false:
    fixes init target
    assumes
        0 < init
        init < target
    shows

\{x::bool\ stream.\ success\ init\ x\ target\ \land\ shd\ x = False\} = \{x::bool\ stream.\ success\ (init-1)\ (stl\ x)\ target\ \land\ shd\ x = False\}
\langle proof\ \rangle
```

1.5 Probability equation

Here we start to analyse the probability of successful random walk. To better understand this part please have a look the elaboration in front of lemma $success_measurable$

 $probability_of_win$ is the function describing possibility of successful random walks with initial number and target number as inputs

```
fun probability-of-win::int \Rightarrow int \Rightarrow ennrealwhere probability-of-win init target = emeasure M {x \in space\ M. success init x target}
```

1.5.1 Successful random walk set is measurable

Preimage of function snth is measurable

```
lemma snth-measurable:
fixes n::nat
shows\bigwedge k. (\lambda w. snth w n) - `\{k\} \in sets M
\langle proof \rangle
```

```
lemma stake-measurable-pre1:

fixes n \le k

assumes length \ k > n

shows stake \ (Suc \ n) \ w = take \ (Suc \ n) \ k \longleftrightarrow stake \ n \ w = take \ n \ k \wedge snth \ w \ n

= nth \ k \ n

\langle proof \rangle
```

```
lemma stake-measurable-pre:

fixes n

shows\land k. length k \ge n \Longrightarrow (stake \ n - `\{k\}) \in sets \ M
```

The preimage of any list over function stake is measurable

 ${f lemma}$ stake-measurable:

```
fixes n \ k

shows(stake n - \{k\}) \in sets \ M

\langle proof \rangle
```

The preimage of any list set over function stake is measurable once the set is finite

```
lemma finite-stake-measurable:
fixes A and n::nat
assumes finite A
shows (stake \ n - 'A) \in sets \ M
\langle proof \rangle
```

The new geom_proc function for list

```
fun geom-proc-list::int \Rightarrow bool list \Rightarrow intwhere
geom-proc-list init [] = init|
geom-proc-list init (x \# xs) = (case \ x \ of \ True \Rightarrow 1|False \Rightarrow -1) + geom-proc-list init xs
```

```
\mathbf{lemma}\ reverse\text{-}construct\text{-}pre\text{:}
      fixes init lengthx y
        shows \bigwedge x::bool\ list.\ lengthx = length\ x \Longrightarrow geom-proc-list\ init\ (x @ [y]) =
geom\text{-}proc\text{-}list\ init\ x + (case\ y\ of\ True \Rightarrow 1|False \Rightarrow -1)
\langle proof \rangle
lemma reverse-construct:
      fixes init x y
   shows geom-proc-list init (x @ [y]) = geom-proc-list init x + (case y of True <math>\Rightarrow 1 | False \Rightarrow 1 | F
-1)
      \langle proof \rangle
lemma success-pre:
     fixes init x target i
      assumes 0 < initinit < target
     shows geom-proc-list init (stake i x) = geom-proc init x i
Any natural number smaller than Inf A doesn't belong to A
lemma not-belong:
      fixes A and a::nat
      assumes \prod A > a
      shows a \notin A
\langle proof \rangle
This is the most important intermediate lemma prepared for lemma success measurable. It clarifiest
\mathbf{lemma}\ success-measurable 2:
      fixes init target and i::nat
     assumes 0 < initinit < target 0 \le i
     shows\{x \in space M. success init x target \land stop-at init x target = i\}
= stake\ i\ -\ (c::bool\ list.\ (\forall\ k < i.\ (geom\-proc\-list\ init\ (take\ k\ c)) \notin \{0, target\}) \land
length \ c = i \land geom\text{-}proc\text{-}list \ init \ c = target\}
\langle proof \rangle
lemma stake-space:stake n ' space M = \{c::bool \ list. \ length \ c = n\}
\langle proof \rangle
Set of all the lists with specific length is finite
lemma finite-length:finite \{c::bool\ list.\ length\ c=n\}
\langle proof \rangle
lemma finite-image:finite \{c::bool\ list.\ (\forall\ k < i.\ (geom-proc-list\ init\ (take\ k\ c))\notin a
\{0, target\} \land length \ c = i \land geom\text{-}proc\text{-}list \ init \ c = target\}
      \langle proof \rangle
Sets of all successful random walk with specific stop is measurable
```

lemma success-measurable3:

```
fixes init and target and i::nat assumes 0 < initinit < target 0 \le i shows\{x \in space \ M. \ success \ init \ x \ target \land stop-at \ init \ x \ target = enat \ i\} \in sets \ M  \langle proof \rangle
```

Any successful random walk must stop at specific position described by natural number

```
lemma success-measurable1:

fixes init target

assumes 0 < initinit < target

shows \{x \in space \ M. \ success \ init \ x \ target\}

= (\bigcup i::nat. \ \{x \in space \ M. \ success \ init \ x \ target \land stop-at \ init \ x \ target = i\})

\langle proof \rangle
```

Here we need to elaborate about this most difficult lemma we've met during this model formalization. lemma success_measurable asserts that successful random walks set under assumption " $0 \le initial number \le target number$ " is measurable set for measure M. On the one hand, since the probability theory has been set up based on the measure theory, every specific set must be proved to be measurable with respect to fixed measure before we calculate the probability of the set, which severely hinders most of scholars and experts from formalizing the security analysis related to the probability since it's extremely difficult to prove why your set is measurable. That is exactly why our endeavor matters to provide the first example to overcome the difficulty. On the other hand, we are willing to briefly explain the way we prove this lemma since it's nontrivial even for pen-and-paper proof. lemma finite stake measurable states that for the function (λw . stake n w) taking the first n steps of random walk, the preimage of a finite sets is measurable for measure M. lemma finite_image states that sets filled with all bool list of fixed length n is finite. lemma success_measurable2 sets up the bijection between successful random walks stopping at fixed step and preimage of successful bool list with identical length. lemma success measurable1 demonstrates that set of successful random walks is countable union of sets of successful random walks stopping at some step. Combining theses 4 lemmas together proves the set of successful random walk is measurable. If you take a closed look at the proofs of these 4 lemmas patiently, you will find it's very hard to finish. Honestly, we will never be able to finish such difficult proofs within one month without the current stochastic process theory library established just in 2021 by Mnacho Echenim, the author of theory infinite coin toss space.

```
lemma success-measurable:

fixes init target

assumes 0 \le initinit \le target

shows \{x \in space \ M. \ success \ init \ x \ target\} \in sets \ M

\langle proof \rangle
```

```
The set of all the random walk with first step True is measurable
\mathbf{lemma}\ success-measurable\text{-}shd:
\{x \in space \ M. \ shd \ x\} \in sets \ M
  \langle proof \rangle
The set of all the random walk with first step False is measurable
{f lemma}\ success-measurable-shd-false:
\{x \in space \ M. \ \neg \ shd \ x\} \in sets \ M
  \langle proof \rangle
lemma success-measurable-final:
  fixes init target
  assumes 0 < init init < target
 shows\{x \in space M. success (init+1) (stl x) target \land shd x\} \in sets M
\langle proof \rangle
         Probability of successful random walk with its first step
          True
lemma semi-goal1:
 fixes init target P
 assumes 0 < initinit \le target \land x. \ P \ x = success \ (init+1) \ (stl \ x) \ target \land shd \ x
 shows emeasure M \{x. P (t \#\# x)\}
= (case\ t\ of\ True \Rightarrow 1|False \Rightarrow 0) * emeasure\ M\ \{x.\ success\ (init+1)\ x\ target\}
\langle proof \rangle
lemma semi-goal21:
 fixes p1 e::real and f
  assumes m = measure-pmf (bernoulli-pmf p1)\bigwedge t. f t = ennreal \ e * (case \ t \ of \ t)
True \Rightarrow 1 | False \Rightarrow 0)
 shows simple-function m f
  \langle proof \rangle
lemma semi-goal21-false:
  fixes p1 e::real and f
  assumes m = measure-pmf (bernoulli-pmf p1)\bigwedge t. f t = ennreal \ e * (case \ t \ of \ t)
True \Rightarrow 0 | False \Rightarrow 1)
  shows simple-function m f
  \langle proof \rangle
lemma sum-rephrase:
  fixes f::ennreal \Rightarrow ennreal and e
  assumes 0 \neq e
 shows sum f \{0,e\} = f 0 + f (e)
```

```
lemma semi-goal 22:
  fixes p1 e::real
  assumes m = measure-pmf (bernoulli-pmf p1)0 \le p1p1 \le 1
\bigwedge t. \ f \ t = ennreal \ e * (case \ t \ of \ True \Rightarrow 1 | False \Rightarrow 0)
  shows integral^S m f = p1 * e
  \langle proof \rangle
lemma semi-goal22-false:
  fixes p1 e::real
  assumes m = measure-pmf (bernoulli-pmf p1)0 \le p1p1 \le 1
\bigwedge t. \ f \ t = ennreal \ e * (case \ t \ of \ True \Rightarrow 0 | \ False \Rightarrow 1)
  shows integral^S m f = (1-p1) * e
  \langle proof \rangle
 lemma semi-goal23:
  fixes p1 e::real and f
  assumes 0 \le p1
          p1 \leq 1
          e > 0
m = measure-pmf (bernoulli-pmf p1)
\bigwedge t. \ f \ t = ennreal \ e * (case \ t \ of \ True \Rightarrow 1 | False \Rightarrow 0)
shows \int_{-\infty}^{\infty} t \cdot (f t) \partial m = integral^{S} m f
  \langle proof \rangle
lemma semi-goal23-false:
  fixes p1 e::real and f
  assumes 0 \le p1
          p1 \leq 1
          e > 0
m = measure-pmf (bernoulli-pmf p1)
\bigwedge t. \ f \ t = ennreal \ e * (case \ t \ of \ True \Rightarrow 0 | \ False \Rightarrow 1)
shows \int f^+ t \cdot (f t) \partial m = integral^S m f
  \langle proof \rangle
lemma semi-goal2:
  fixes p1 e::real and f
  assumes 0 \le p1
          p1 \leq 1
          e \ge 0
m = measure-pmf (bernoulli-pmf p1)
\bigwedge t. \ f \ t = ennreal \ e * (case \ t \ of \ True \Rightarrow 1 | False \Rightarrow 0)
shows \int + t \cdot (f t) \partial m = p1 * e
```

 $\langle proof \rangle$

```
\langle proof \rangle
\mathbf{lemma}\ semi\text{-}goal \textit{2-}false:
  fixes p1 e::real and f
  assumes 0 \le p1
           p1 \leq 1
           e \geq 0
m = measure-pmf (bernoulli-pmf p1)
\bigwedge t. \ f \ t = ennreal \ e * (case \ t \ of \ True \Rightarrow 0 | \ False \Rightarrow 1)
shows \int_{-\infty}^{\infty} t \cdot (f t) \partial m = (1-p1) \cdot e
\langle proof \rangle
lemma semi-goal2-final:
  fixes p1::real and e::ennreal and f
  assumes 0 < p1
           p1 \leq 1
           e \neq top
m = measure-pmf (bernoulli-pmf p1)
\bigwedge t. \ f \ t = e * (case \ t \ of \ True \Rightarrow 1 | False \Rightarrow 0)
shows \int + t \cdot (f t) \partial m = p1 * e
\langle proof \rangle
\mathbf{lemma}\ semi\text{-}goal \textit{2-}final\text{-}false:
  fixes p1::real and e::ennreal and f
  assumes 0 \le p1
           p1 \leq 1
           e \neq top
m = measure-pmf (bernoulli-pmf p1)
\bigwedge t. \ f \ t = e * (case \ t \ of \ True \Rightarrow 0 | \ False \Rightarrow 1)
shows \int_{-\infty}^{\infty} t \cdot (f t) \partial m = (1 - p1) \cdot e
\langle proof \rangle
lemma fun-description-pre:
  fixes init target t
  assumes \theta < initinit < target
  shows
emeasure M {x \in space M. t \# \# x \in \{x \in space M. success (init+1) (stl x) target}
\land shd x\}
= (case\ t\ of\ True \Rightarrow 1|False \Rightarrow 0) * (emeasure\ M\ \{x \in space\ M.\ success\ (init+1)\}
(x) target\})
\langle proof \rangle
lemma fun-description-pre-false:
  fixes init target t
  assumes 0 < initinit < target
```

shows

```
emeasure M \{x \in space \ M. \ t \#\# x \in \{x \in space \ M. \ success \ (init-1) \ (stl \ x) \ target \land \neg \ shd \ x\}\}
= (case \ t \ of \ True \Rightarrow \ 0 | False \Rightarrow \ 1) * (emeasure \ M \ \{x \in space \ M. \ success \ (init-1) \ (x) \ target\})
\langle proof \rangle
```

 $\mathbf{term} \, emeasure\text{-}stream\text{-}space$

The lemma <code>semi_goal_true</code> is the second difficulty we've overcome during the model formalization. It asserts that probability of sets of successful random walk with first step True is equal to probability of sets of random walk times probability of sets of successful random walk with initial number plus 1. Thanks to the lemma <code>emeasure_stream_space</code> provided by Mnacho Echenim, the author of <code>infinite_coin_toss_space</code>, we could finally use the integral rather than tediously break down the countable product to calculate the probability

```
lemma semi-goal-true:

fixes init target

assumes 0 < initinit < target

shows emeasure M {x \in space \ M. \ success \ (init+1) \ (stl \ x) \ target \land shd \ x}

= emeasure M {x \in space \ M. \ shd \ x} * emeasure M {x \in space \ M. \ success \ (init+1) \ (x) \ target}

\langle proof \rangle
```

```
lemma semi-goal-false:

fixes init target

assumes 0 < initinit < target

shows emeasure M {x \in space \ M. \ success \ (init-1) \ (stl \ x) \ target \land \neg \ shd \ x}

= emeasure M {x \in space \ M. \ \neg \ shd \ x} * emeasure M {x \in space \ M. \ success \ (init-1) \ (x) \ target}

\langle proof \rangle
```

1.5.3 Final goal: establish the recursive probability equation

The final probability equation we want to formalize:

$$P_n = pP_{n+1} + (1-p)P_{n-1}$$

```
lemma Recursive-probability-equation:

fixes init target

assumes 0 < init init < target

showsprobability-of-win init target = p * (probability-of-win (init + 1) target) + (1 - p) * (probability-of-win (init - 1) target)

<math>\langle proof \rangle
```

end end